

Introduction to Tensors

To understand tensors, one needs to understand three kinds of vector spaces and be able to switch back and forth between them: **abstract** vector spaces, vector spaces of **linear functionals**, and vector spaces of **matrices**.

Tensors are defined in terms **linear functionals**. Tensors are manipulated as **abstract** symbols. Computations among tensors are often carried out using **matrices**.

For this reason, we begin by providing the requisite background in **abstract** vector spaces, **linear functionals**, and **matrices**.

1 Vector spaces

Definition An **n -dimensional vector space \mathcal{V} over a field \mathbf{F}** is an **abstract** collection of objects called **vectors** that is closed under addition and scalar multiplication:

$$a\mathbf{v} + b\mathbf{w} \in \mathcal{V} \text{ if } a \text{ and } b \text{ are scalars and } \mathbf{v} \text{ and } \mathbf{w} \text{ are vectors}$$

The vector space definition also includes several rules that ensure proper behavior of addition and scalar multiplication, such as forming an Abelian group under addition, that will not be listed here.

Definition **Scalars** are simply elements of \mathbf{F} . For quantum mechanics, the field is the complex numbers, \mathbb{C} . For spacetime, \mathbb{R} is used.

Definition A **basis** is a linearly independent set of vectors that span the vector space. Let $\mathcal{B} = \{\mathbf{e}_{(\mu)}\} = \{\mathbf{e}_{(1)}, \dots, \mathbf{e}_{(n)}\}$ denote a basis for \mathcal{V} . Vectors will be displayed in bold face. Parentheses will be used on subscripts to emphasize that $\mathbf{e}_{(\mu)}$ are vectors as opposed to coefficients of vectors.

Definition Every vector in \mathcal{V} can be uniquely expressed as a **linear combination** of basis vectors:

$$\mathbf{V} = V^\mu \mathbf{e}_{(\mu)} = V^1 \mathbf{e}_{(1)} + \dots + V^n \mathbf{e}_{(n)}.$$

This expression uses the Einstein summation convention to add products of terms having matching upper and lower Greek indices. The lack of parentheses on the superscripts of V emphasizes that they are components and not vectors. Components will not be put in bold face.

Notation To represent the abstract vector space \mathcal{V} as a vector space of column vectors we identify the basis vectors $\mathbf{e}_{(\mu)}$ as column vectors:

$$\mathbf{e}_{(\mu)} = \begin{pmatrix} 0 \\ \vdots \\ 1^\mu \\ \vdots \\ 0 \end{pmatrix}.$$

Every vector in \mathcal{V} can be expressed as a linear combination of the basis vectors:

$$\mathbf{V} = V^\mu \mathbf{e}_{(\mu)} = \begin{pmatrix} V^1 \\ \vdots \\ V^\mu \\ \vdots \\ V^n \end{pmatrix}.$$

The vector $\mathbf{V} = V^\mu \mathbf{e}_{(\mu)}$ is sometimes loosely referred to by its component V^μ .

This is done because $\mathbf{e}_{(\mu)}$ are only place-holder zeros and ones; the information content is carried by the components.

Definition A **covector** is a row vector $\omega = (\omega_1 \ \dots \ \omega_n)$. As with vectors, the covector ω is sometimes loosely referred to as ω_μ . Note that if \mathbf{V} is a vector, then its **transpose**, \mathbf{V}^T , is a covector.

Notation Let \mathcal{V}^* be an n -dimensional vector space of covectors, and $\mathcal{B}^* = \{\varepsilon^{(\nu)}\} = \{\varepsilon^{(1)}, \dots, \varepsilon^{(n)}\}$ a basis where $\varepsilon^{(\nu)} = (0 \ \dots \ 1_\nu \ \dots \ 0)$. \mathcal{V}^* is called a **covector space**. Every covector can be expressed as a linear combination $\omega = \omega_\nu \varepsilon^{(\nu)} = (\omega_1 \ \dots \ \omega_n)$ of basis covectors.

Theorem The product of any row vector with any column vector is a scalar:

$$\omega \mathbf{V} = (\omega_1 \ \dots \ \omega_n) \begin{pmatrix} V^1 \\ \vdots \\ V^n \end{pmatrix} = \omega_\mu V^\mu \in \mathbb{F}. \quad (\text{i})$$

Definition A **functional** is a function whose range is the field, \mathbf{F} . A covector ω can be regarded as a **linear functional** by the formula:

$$\omega : \mathcal{V} \rightarrow \mathbf{F} : \omega(\mathbf{V}) = \omega \mathbf{V}.$$

Linear means that the vector space operations of addition and scalar multiplication are preserved:

$$\omega(a\mathbf{V} + b\mathbf{W}) = a\omega(\mathbf{V}) + b\omega(\mathbf{W}).$$

As illustrated, we use boldface for covectors like ω but non-boldface for functions like ω .

Construction Since the linear functionals ω can be mapped 1-1 with the covectors ω , it should not be surprising that the functionals can be structured to form a vector space. We define addition of functions:

$$(\omega + \xi)(\mathbf{V}) \equiv \omega(\mathbf{V}) + \xi(\mathbf{V})$$

and scalar multiplication:

$$(a\omega)(\mathbf{V}) = a[\omega(\mathbf{V})].$$

Then the function $\omega + \xi$ corresponds to the covector $\omega + \xi$ and the function $a\omega$ corresponds to the covector $a\omega$. The additive identity is the zero function, $-\omega$ is the additive inverse of ω , and all the remaining vector space rules for addition and scalar multiplication are easily verified. We have proven the following theorem.

Theorem The mapping $\omega \mapsto \omega$ is an **isomorphism**. That is, it is a 1-1 mapping that preserves the vector space structure of addition and scalar multiplication.

If we have a functional, ω , we are free to switch on-the-fly and consider it a covector, ω , and vice-versa.

Definition We have shown that \mathcal{V}^* can thus be viewed from three perspectives:

(1) **abstract** vector space, (2) vector space of **linear functionals**, and (3) vector space of **row vectors** (i.e., $1 \times n$ **matrices**). The abstract perspective simply means viewing members of \mathcal{V}^* as abstract vectors expressed in terms of a basis:

$\omega = \omega_\nu \varepsilon^{(\nu)}$. In the functional perspective, \mathcal{V}^* is called the **dual space of \mathcal{V}** , and its elements ω are linear transformations from \mathcal{V} to \mathbf{F} , known as **dual vectors**.

Observe that in the vector space perspective we have

$$\varepsilon^{(\nu)} \mathbf{e}_{(\mu)} = \begin{pmatrix} \dots & 1_\nu & \dots \end{pmatrix} \begin{pmatrix} \vdots \\ 1^\mu \\ \vdots \end{pmatrix} = \delta_\mu^\nu.$$

To be consistent, when we work in either the linear functional perspective or the abstract perspective, we will require that the bases of \mathcal{V} and \mathcal{V}^* satisfy

$\varepsilon^{(\nu)} \mathbf{e}_{(\mu)} = \delta_\mu^\nu$. An aside: since $\varepsilon^{(i)} \mathbf{e}_{(i)} = 1$, we can define the inverse, $\frac{1}{\mathbf{e}_{(\mu)}} \equiv \varepsilon^{(\mu)}$. We

see that a subscript in the denominator is effectively a superscripted variable.

In most books the terms "covector", "linear functional", "dual vector", and "1-form" are used interchangeably. There are actually subtle distinctions between them. For the record, we distinguish them now.

1. A **covector** refers to a row vector and belongs to the matrix-algebra perspective.
2. A **linear functional** is a linear map from a vector space to a field \mathbf{F} and belongs to the linear functional perspective.
3. When the set of linear functionals is considered as a vector space, the vector space is known as the **dual space** and the linear functionals are referred to as **dual vectors**. They are "vectors" because we are considering them as part of a vector space of linear functionals.
4. The term 1-form is most often used in differential geometry, which plays a role in general relativity. A **manifold** is a topological space that is locally homeomorphic to Euclidean space. A **vector field**, or just a **field**, is a set of vectors, one for every point on a manifold. A **1-form** is a vector field where the vectors are linear functionals. Somewhat confusingly, the term "1-form" is loosely and frequently used also to refer to individual linear functionals in the field. In physics, the term 1-form is most often used when the linear functionals are differentials, ∂f . In tensor theory, a 1-form refers to a $(0, 1)$ tensor, soon to be defined, which is simply a dual vector.

Definition One can define \mathcal{V}^{**} , the dual space of the dual space \mathcal{V}^* . This space is isomorphic to \mathcal{V} via the natural 1-1 mapping $V^{**}(\omega) \equiv \omega(V)$, where $V^{**} \in \mathcal{V}$.

In this way, \mathcal{V} can be regarded as the dual space of \mathcal{V}^* , just as \mathcal{V}^* is the dual space of \mathcal{V} .

A way to understand the importance of this is as follows. Just as a row vector can be viewed as a linear functional when it acts on a column vector, a column vector can be viewed as a linear functional when it acts on a row vector. Tensors are defined as linear functionals, so the perspective that row and column vectors are duals of each other plays a part.

We have learned how to go from row and column vectors to linear functionals. We also need to perform this process in reverse: identify a column vector that corresponds to a linear functional. Fortunately, this is very easy.

Construction Let $\{T : \mathcal{V} \rightarrow \mathbf{F}\}$ be a vector space of linear functionals and let $\{\mathbf{e}_{(\mu)}\} = \{\mathbf{e}_{(1)}, \dots, \mathbf{e}_{(n)}\}$ be a basis, where each $\mathbf{e}_{(\mu)} : \mathcal{V} \rightarrow \mathbf{F}$ is, of course, a linear functional. First, express a linear functional T abstractly in terms of the basis

elements: $T = T^\mu \mathbf{e}_{(\mu)}$. Next, identify the column vector $\begin{pmatrix} 0 \\ \vdots \\ 1^\mu \\ \vdots \\ 0 \end{pmatrix}$ with the functional $\mathbf{e}_{(\mu)}$. The collection of these column vectors forms a basis for the vector space of length n column vectors: $\mathbf{T} = T^\mu \mathbf{e}_{(\mu)} = \begin{pmatrix} T^1 \\ \vdots \\ T^n \end{pmatrix}$ and identify T with \mathbf{T} . We are free to think of T either as a linear functional or as \mathbf{T} , a column vector. We will be careful to use boldface for matrix objects and non-boldface for linear operators.

We have actually just proved that every n -dimensional vector space is isomorphic to the vector space of length n column vectors. This amounts to a very important theorem.

Theorem All n -dimensional vector spaces are isomorphic.

In this sense, there is just a single vector space of each finite dimension. For example, $n \times n$ matrices, n^2 -dimensional linear functionals, and \mathbb{R}^{2n} are not the same, but in terms of just their vector space properties, they are identical.

Definition The **product $\mathcal{V} \times \mathcal{W}$ of two vector spaces** is defined to be the vector space of pairs (\mathbf{V}, \mathbf{W}) where $\mathbf{V} \in \mathcal{V}$ and $\mathbf{W} \in \mathcal{W}$. Addition and scalar multiplication mimics addition and scalar multiplication of (x, y) pairs in \mathbb{R}^2 :

$$(\mathbf{V}, \mathbf{W}) + (\mathbf{X}, \mathbf{Y}) = (\mathbf{V} + \mathbf{X}, \mathbf{W} + \mathbf{Y}) \text{ and } \alpha(\mathbf{V}, \mathbf{W}) = (\alpha \mathbf{V}, \alpha \mathbf{W}).$$

As we have seen, every linear functional can be considered to be a column vector. A linear functional defined on a product of vector spaces can also be viewed as a matrix. This identification is the critical step (9) in Section 4 where we develop rank $(1, 1)$ tensors. We carry out that step now.

Construction Suppose \mathcal{V} and \mathcal{W} are n -dimensional vector spaces, \mathcal{V}^* the dual space of \mathcal{V} , $\{T : \mathcal{V}^* \times \mathcal{W} \rightarrow \mathbf{F}\}$ a vector space of linear functionals, and $\{\mathbf{e}_{(\mu)}^{(\nu)}\}$ a basis for $\{T\}$. Let $T = T_\nu^\mu \mathbf{e}_{(\mu)}^{(\nu)}$. We can identify the function $\mathbf{e}_{(\mu)}^{(\nu)}$ with the matrix

$$\mathbf{e}_{(\mu)}^{(\nu)} \equiv \begin{pmatrix} 0 & \cdots & 0 & \cdots & 0 \\ \vdots & & \vdots & & \vdots \\ 0 & \cdots & 1_\nu^\mu & \cdots & 0 \\ \vdots & & \vdots & & \vdots \\ 0 & \cdots & 0 & \cdots & 0 \end{pmatrix} \quad (\text{ii})$$

that has a '1' in (row μ , column ν). The collection $\{\mathbf{e}_{(\mu)}^{(\nu)}\}$ forms a basis for the vector space of nxn matrices, and so we can identify the linear functional T with the matrix $\mathbf{T} = T_\nu^\mu \mathbf{e}_{(\mu)}^{(\nu)}$.

Notice above that the coefficient T_ν^μ is the same for both the matrix \mathbf{T} and the function T . **We will interchangeably refer to the abstract symbol T_ν^μ as a linear functional or a matrix.**

Except for quantum mechanics, in the literature all three of T_ν^μ , $T_{\mu\nu}$, and $T^{\mu\nu}$ are usually treated as matrices. We will treat only the first one as a matrix. We prefer to represent the middle one as a $(1 \times n^2)$ row vector, and the 3rd one as a $(n^2 \times 1)$ column vector. Examples will be provided in Section 2. Because the vector spaces of (nxn) matrices, $(1 \times n^2)$ row vectors, and $(n^2 \times 1)$ column vectors have the same vector space dimension n^2 , they are isomorphic to each other. So, treating all three of T_ν^μ , $T_{\mu\nu}$, and $T^{\mu\nu}$ as matrices is also legitimate.

It should be mentioned that linear operators that have 3 or more indices can be represented not only as row vectors, column vectors, and 2-dimensional matrices, but also 3-dimensional matrices, and more. We will treat all linear operators as either row vectors (e.g., $T_{\mu\nu o}$), column vectors (e.g., $T^{\mu\nu o\rho}$), or 2-D matrices (e.g., $T_{\nu o \sigma}^{\mu\rho}$).

2 Tensor Products of Matrices

Definition Equation (i), $\omega \mathbf{V} = \omega_\mu V^\mu$, is an example of an **inner product**, a sum of products of matching elements. Another kind of product is an **outer product**, where every element of \mathbf{V} is multiplied by every element of ω :

$$\begin{pmatrix} V^1 \\ \vdots \\ V^n \end{pmatrix} \begin{pmatrix} \omega_1 & \cdots & \omega_n \end{pmatrix} = \begin{pmatrix} V^1 \omega_1 & \cdots & V^1 \omega_n \\ \vdots & & \vdots \\ V^n \omega_1 & \cdots & V^n \omega_n \end{pmatrix}$$

Definition Matrix multiplication is not defined to include outer products. Because it is useful for tensors, there is an outer product operation. It is called **tensor product**, and it is denoted by \otimes . We proceed to define $\mathbf{A} \otimes \mathbf{B}$ where \mathbf{A} and \mathbf{B} are any matrix objects: scalars, row vectors, column vectors, or matrices. We first define \otimes for a scalar k and a matrix object \mathbf{A} , where \mathbf{A} is either a matrix, a row vector, or a column vector.

$$k \otimes \mathbf{A} = \mathbf{A} \otimes k \equiv k\mathbf{A} \quad (\text{iii})$$

For example, $k \otimes \begin{pmatrix} a & b \\ c & d \end{pmatrix} = k \begin{pmatrix} a & b \\ c & d \end{pmatrix} = \begin{pmatrix} ka & kb \\ kc & kd \end{pmatrix}$.

We complete the definition by defining \otimes for two non-scalar matrix objects:

$$\mathbf{A} \otimes \mathbf{B} \equiv \begin{pmatrix} A_{11} \otimes \mathbf{B} & \cdots & A_{1n} \otimes \mathbf{B} \\ \vdots & & \vdots \\ A_{n1} \otimes \mathbf{B} & \cdots & A_{nn} \otimes \mathbf{B} \end{pmatrix} \quad (\text{iv})$$

Note that this bootstrap definition depends on (iii) because, for example, A_{11} is a scalar. We refer to this as a **pattern definition** because the operation is easily carried out by following the pattern (iv).

Example 1: Two matrices:

$$\mathbf{A} = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{21} \end{pmatrix}, \quad \mathbf{B} = \begin{pmatrix} b_{11} & b_{12} \\ b_{21} & b_{21} \end{pmatrix}$$

$$\mathbf{A} \otimes \mathbf{B} \stackrel{(iv)}{=} \left(\begin{array}{cc} \mathbf{a}_{11} \otimes \mathbf{B} & \mathbf{a}_{12} \otimes \mathbf{B} \\ \mathbf{a}_{21} \otimes \mathbf{B} & \mathbf{a}_{21} \otimes \mathbf{B} \end{array} \right) = \left(\begin{array}{cc|cc} \mathbf{a}_{11} \otimes \mathbf{b}_{11} & \mathbf{a}_{11} \otimes \mathbf{b}_{12} & | & \mathbf{a}_{12} \otimes \mathbf{b}_{11} & \mathbf{a}_{12} \otimes \mathbf{b}_{12} \\ \mathbf{a}_{11} \otimes \mathbf{b}_{21} & \mathbf{a}_{11} \otimes \mathbf{b}_{22} & | & \mathbf{a}_{12} \otimes \mathbf{b}_{21} & \mathbf{a}_{12} \otimes \mathbf{b}_{22} \\ \hline - & - & + & - & - \\ \mathbf{a}_{21} \otimes \mathbf{b}_{11} & \mathbf{a}_{21} \otimes \mathbf{b}_{12} & | & \mathbf{a}_{22} \otimes \mathbf{b}_{11} & \mathbf{a}_{22} \otimes \mathbf{b}_{12} \\ \mathbf{a}_{21} \otimes \mathbf{b}_{21} & \mathbf{a}_{21} \otimes \mathbf{b}_{22} & | & \mathbf{a}_{22} \otimes \mathbf{b}_{21} & \mathbf{a}_{22} \otimes \mathbf{b}_{22} \end{array} \right)$$

Example 2: Column vector and a row vector:

$$\mathbf{V} \otimes \omega \stackrel{(iv)}{=} \left(\begin{array}{c} \mathbf{V}^1 \otimes \omega \\ \vdots \\ \mathbf{V}^n \otimes \omega \end{array} \right) = \left(\begin{array}{c} \mathbf{V}^1 \otimes \left(\begin{array}{ccc} \omega_1 & \dots & \omega_n \end{array} \right) \\ \vdots \\ \mathbf{V}^n \otimes \left(\begin{array}{ccc} \omega_1 & \dots & \omega_n \end{array} \right) \end{array} \right) \\ \equiv \left(\begin{array}{ccc} \mathbf{V}^1 \otimes \omega_1 & \dots & \mathbf{V}^1 \otimes \omega_n \\ \vdots & & \vdots \\ \mathbf{V}^n \otimes \omega_1 & \dots & \mathbf{V}^n \otimes \omega_n \end{array} \right)$$

or

$$= \left(\begin{array}{ccc} (\mathbf{V}^1, \omega_1) & \dots & (\mathbf{V}^1, \omega_n) \\ \vdots & & \vdots \\ (\mathbf{V}^n, \omega_1) & \dots & (\mathbf{V}^n, \omega_n) \end{array} \right).$$

The tensor symbols or the ordered pairs inside the matrices are simply reminders not to combine the scalars \mathbf{V}^μ and ω_μ .

As a particular example of this, we can re-generate equation (ii) as the tensor product of a column basis vector and a row basis vector:

$$\mathbf{e}_{(\mu)} \otimes \varepsilon^{(\nu)} = \left(\begin{array}{c} \vdots \\ \mathbf{1}^\mu \\ \vdots \end{array} \right) \otimes \left(\begin{array}{ccc} \dots & \mathbf{1}_\nu & \dots \end{array} \right) = \left(\begin{array}{ccc} \dots & \vdots & \dots \\ \dots & \mathbf{1}_\nu^\mu & \dots \\ \vdots & & \dots \end{array} \right) \quad (v)$$

Example 3: Two row vectors:

$$\begin{pmatrix} \omega_1 & \cdots & \omega_n \end{pmatrix} \otimes \begin{pmatrix} \omega_1 & \cdots & \omega_n \end{pmatrix} \stackrel{(iv)}{=} \begin{pmatrix} \omega_1 \begin{pmatrix} \omega_1 & \cdots & \omega_n \end{pmatrix} & \cdots & \omega_n \begin{pmatrix} \omega_1 & \cdots & \omega_n \end{pmatrix} \end{pmatrix}$$
$$\begin{pmatrix} \omega_1\omega_1 & \cdots & \omega_1\omega_n & \omega_2\omega_1 & \cdots & \omega_2\omega_n & \cdots & \omega_n\omega_1 & \cdots & \omega_n\omega_n \end{pmatrix}$$

When ω is an n covector, $\omega \otimes \omega$ is an n^2 covector, just as $\mathbf{V} \otimes \omega$, above, is an $n \times n$ matrix. Similarly, the tensor product of two length n column vectors would be expressed as a length n^2 column vector.

3 Tensors

We now have the machinery to define tensors. Though tensors can be defined for infinite-dimensional vector spaces, we will restrict our scope to finite-dimensional vector spaces. Tensors are defined as [linear functionals](#). Identifying them with [matrix objects](#) is performed in the next section.

Definition Let $\mathcal{V}, \dots, \mathcal{W}$ be finite-dimensional vector spaces, and $\mathcal{Y}^*, \dots, \mathcal{Z}^*$ be finite-dimensional dual spaces. Let k and ℓ be non-negative integers. A **rank (k, ℓ) tensor** is a multilinear map T from a product of dual vector spaces and vector spaces to \mathbf{F} :

$$T : \underbrace{\mathcal{Y}^* \times \cdots \times \mathcal{Z}^*}_{k \text{ terms}} \times \underbrace{\mathcal{V} \times \cdots \times \mathcal{W}}_{\ell \text{ terms}} \rightarrow \mathbf{F}. \quad (1)$$

Multilinear means the tensor T acts linearly in each of its arguments. For example, for a $(1, 1)$ tensor,

$$\begin{aligned} & T\left(a_1 \omega^{(1)} + a_2 \omega^{(2)}, b_1 \mathbf{V}_{(1)} + b_2 \mathbf{V}_{(2)}\right) \\ &= a_1 b_1 T\left(\omega^{(1)}, \mathbf{V}_{(1)}\right) + a_1 b_2 T\left(\omega^{(1)}, \mathbf{V}_{(2)}\right) + a_2 b_1 T\left(\omega^{(2)}, \mathbf{V}_{(1)}\right) + a_2 b_2 T\left(\omega^{(2)}, \mathbf{V}_{(2)}\right). \end{aligned}$$

The upper indices are called **contravariant indices** and the lower ones are called **covariant indices**.

As with linear functionals, addition and scalar multiplication of tensor functions are defined naturally, resulting that [the set of \$\(k, \ell\)\$ tensors form a vector space under addition and scalar multiplication](#). In particular,

$$\begin{aligned} & [aT + bS]\left(\omega^{(1)}, \dots, \omega^{(k)}, V_{(1)}, \dots, V_{(\ell)}\right) \\ & \equiv a\left[T\left(\omega^{(1)}, \dots, \omega^{(k)}, V_{(1)}, \dots, V_{(\ell)}\right)\right] + b\left[S\left(\omega^{(1)}, \dots, \omega^{(k)}, V_{(1)}, \dots, V_{(\ell)}\right)\right]. \end{aligned}$$

Just as a linear functional T can be identified with the matrix $\mathbf{T} = T_{\nu}^{\mu} \mathbf{e}_{(\mu)}^{(\nu)}$, we will show that a tensor is isomorphic to a matrix (or, sometimes, a column vector or row vector.)

To keep development simple, we begin with the $(1, 1)$ tensors.

4 Rank (1,1) Tensors

Tensors and their properties are developed in 17 steps, listed here all in one place. It is suggested that this page be printed out for quick reference as the steps are explained, one-by-one. This may look overwhelming but most of the steps are simply definitions.

- (1) $\mathcal{V} \otimes \mathcal{W}^* \equiv \{T : \mathcal{V}^* \times \mathcal{W} \rightarrow \mathbf{F} : T \text{ is bilinear}\}$
- (2) $\mathcal{B}_v = \left\{ \mathbf{e}_{(\mu)} \right\}_{\mu=1}^n, \mathcal{B}_w = \left\{ \mathbf{f}_{(\rho)} \right\}_{\rho=1}^m, \mathcal{B}_{v^*} = \left\{ \boldsymbol{\varepsilon}^{(\nu)} \right\}_{\nu=1}^n, \text{ and } \mathcal{B}_{w^*} = \left\{ \boldsymbol{\varphi}^{(\sigma)} \right\}_{\sigma=1}^m$
- (3) $\mathbf{V} = V^\mu \mathbf{e}_{(\mu)} \in \mathcal{V}, \mathbf{W} = W^\rho \mathbf{f}_{(\rho)} \in \mathcal{W}, \boldsymbol{\omega} = \omega_\nu \boldsymbol{\varepsilon}^{(\nu)} \in \mathcal{V}^*, \text{ and } \boldsymbol{\xi} = \xi_\sigma \boldsymbol{\varphi}^{(\sigma)} \in \mathcal{W}^*$
- (4) $\mathbf{e}_\mu^\nu \left(\boldsymbol{\varepsilon}^{(\sigma)}, \mathbf{f}_{(\rho)} \right) \equiv \delta_\mu^\sigma \delta_\rho^\nu$
- (5) $\mathbf{e}_\mu^\nu : \mathcal{V}^* \times \mathcal{W} : \mathbf{e}_\mu^\nu (\boldsymbol{\omega}, \mathbf{W}) = \omega_\mu W^\nu$
- (6) $\mathcal{B}_{v \otimes w^*} = \left\{ \mathbf{e}_\mu^\nu \right\}_{\mu=1}^n \left\{ \boldsymbol{\varepsilon}^{(\nu)} \right\}_{\nu=1}^m$
- (7) $T_\nu^\mu \equiv T \left(\boldsymbol{\varepsilon}^{(\mu)}, \mathbf{f}_{(\nu)} \right)$
- (8) $T = T_\nu^\mu \mathbf{e}_\mu^\nu$
- (9) There is a 1-1 map between the functionals \mathbf{e}_μ^ν and the matrices $\mathbf{e}_{(\mu)}^{(\nu)}$
- (10) $\mathbf{V} \otimes \boldsymbol{\xi} \equiv V^\mu \xi_\nu \mathbf{e}_{(\mu)}^{(\nu)}$
- (11) $\mathbf{e}_{(\mu)} \otimes \boldsymbol{\varphi}^{(\nu)} = \mathbf{e}_{(\mu)}^{(\nu)}$
- (12) $\mathbf{T} = T_\nu^\mu \mathbf{e}_{(\mu)} \otimes \boldsymbol{\varphi}^{(\nu)}$
- (13) $\mathcal{V} \otimes \mathcal{W}^* = \left\{ T_\nu^\mu \mathbf{e}_{(\mu)} \otimes \boldsymbol{\varphi}^{(\nu)} \right\}$
- (14) $T(\boldsymbol{\omega}, \mathbf{W}) = T_\nu^\mu \omega_\mu W^\nu$
- (15) $T(\boldsymbol{\omega}, \mathbf{W}) = \boldsymbol{\omega} \mathbf{T} \mathbf{W}$
- (16) $\mathbf{T} = \mathbf{T}_\omega \otimes \mathbf{T}^w$
- (17) If $\mathbf{R} = R_\nu^\mu \mathbf{e}_{(\mu)} \otimes \boldsymbol{\varphi}^{(\nu)} \in \mathcal{V} \otimes \mathcal{W}^*$ and $\mathbf{S} = S_\rho^\sigma \mathbf{f}_{(\sigma)} \otimes \boldsymbol{\varepsilon}^{(\rho)} \in \mathcal{W} \otimes \mathcal{V}^*$, define a $(2k, 2\ell)$ tensor in $\mathcal{V} \otimes \mathcal{W} \otimes \mathcal{W}^* \otimes \mathcal{V}^*$ by

$$\begin{aligned} \mathbf{T} &= \mathbf{R} \otimes \mathbf{S} = R_\nu^\mu \mathbf{e}_{(\mu)} \otimes \boldsymbol{\varphi}^{(\nu)} \otimes S_\rho^\sigma \mathbf{f}_{(\sigma)} \otimes \boldsymbol{\varepsilon}^{(\rho)} \\ &\equiv R_\nu^\mu S_\rho^\sigma \mathbf{e}_{(\mu)} \otimes \mathbf{f}_{(\sigma)} \otimes \boldsymbol{\varphi}^{(\nu)} \otimes \boldsymbol{\varepsilon}^{(\rho)} = T_{\nu\rho}^{\mu\sigma} \mathbf{e}_{(\mu)} \otimes \mathbf{f}_{(\sigma)} \otimes \boldsymbol{\varphi}^{(\nu)} \otimes \boldsymbol{\varepsilon}^{(\rho)} \end{aligned}$$

where $T_{\nu\rho}^{\mu\sigma} = R_\nu^\mu S_\rho^\sigma$.

Step 1

This is a repeat of the tensor definition (1) given on page 10 but tailored for just one covector space and one vector space.

$$T : \mathcal{V}^* \times \mathcal{W} \rightarrow \mathbf{F} : T \text{ is bilinear.} \quad (1)$$

$\mathcal{V} \otimes \mathcal{W}^*$ is defined as the collection of all such T s.

Note 1. Note the use of both $\mathcal{V} \otimes \mathcal{W}^*$ and $\mathcal{V}^* \otimes \mathcal{W}$. As we learned in Section 1, a *vector* acts as a function on a covector, and a covector acts as a function on a *vector*.

Note 2. We are using the **tensor symbol**, \otimes , in anticipation that T will be shown to be a tensor. An actual tensor product will not be introduced until step (10). Until then, T , and also $\mathcal{V} \otimes \mathcal{W}$, should be regarded only as **linear functionals**. In (16) we will explicitly express T as a **tensor product** of vectors, $T = \mathbf{V} \otimes \mathbf{W}$. When that happens, it will justify use of the tensor symbol in the vector space name on the LHS of (1).

Step 2

This establishes notation for the basis vectors. We need four basis sets, one each for \mathcal{V} , \mathcal{W} , \mathcal{V}^* , and \mathcal{W}^* .

$$\mathcal{B}_v = \left\{ \mathbf{e}_{(\mu)} \right\}_{\mu=1}^n, \quad \mathcal{B}_w = \left\{ \mathbf{f}_{(\rho)} \right\}_{\rho=1}^m, \quad \mathcal{B}_{v^*} = \left\{ \boldsymbol{\varepsilon}^{(\nu)} \right\}_{\nu=1}^n, \quad \text{and} \quad \mathcal{B}_{w^*} = \left\{ \boldsymbol{\varphi}^{(\sigma)} \right\}_{\sigma=1}^m. \quad (2)$$

Step 3

Using the four bases, we can express vectors belonging to \mathcal{V} , \mathcal{W} , \mathcal{V}^* , and \mathcal{W}^* :

$$\mathbf{V} = V^\mu \mathbf{e}_{(\mu)} \in \mathcal{V}, \quad \mathbf{W} = W^\rho \mathbf{f}_{(\rho)} \in \mathcal{W}, \quad \boldsymbol{\omega} = \omega_\nu \boldsymbol{\varepsilon}^{(\nu)} \in \mathcal{V}^*, \quad \boldsymbol{\xi} = \xi_\sigma \boldsymbol{\varphi}^{(\sigma)} \in \mathcal{W}^*. \quad (3)$$

We use boldface Latin for vectors and boldface Greek for covectors.

Step 4

$$\mathbf{e}_\mu^\nu \left(\boldsymbol{\varepsilon}^{(\sigma)}, \mathbf{f}_{(\rho)} \right) \equiv \delta_\mu^\sigma \delta_\rho^\nu \quad (4)$$

is the definition of \mathbf{e}_μ^ν , a member of the basis for $\mathcal{V} \otimes \mathcal{W}^*$. Two things to notice.

- (a) it is defined as a **linear functional**, and (b) it is defined here only on basis pairs.

Step 5

This is the formula for extending \mathbf{e}_μ^ν to all covector-vector pairs:

$$\mathbf{e}_\mu^\nu (\boldsymbol{\omega}, \mathbf{W}) \stackrel{(3)}{=} \mathbf{e}_\mu^\nu \left(\omega_\sigma \boldsymbol{\varepsilon}^{(\sigma)}, W^\rho \mathbf{f}_{(\rho)} \right) \stackrel{(4)}{=} \omega_\sigma W^\rho \delta_\mu^\sigma \delta_\rho^\nu = \omega_\mu W^\nu \quad \checkmark \quad (5)$$

In defining e_ν^μ , there are reasons why some books slant the indices NW to SE or SW to NE:

1. Without slanting, one cannot immediately tell in (5) which index operates on ω and which operates on \mathbf{W} . In fact, μ acts on ω and ν acts on \mathbf{W} . Correct slanting would be from SW to NE.
2. If an operation involves, for example, a tensor transpose, the slant could identify the tensor versus its transpose. Simply reversing indices to indicate the transpose could be confusing since the indices are dummies and we are free to interchange them at any time.

Nonetheless, we have chosen to keep indices vertical because the slant takes up a lot of space when we get to multiple superscripts and subscripts.

To prove that $e_\mu^\nu \in \mathcal{V} \otimes \mathcal{W}^*$, we must show that e_μ^ν is bilinear:

Suppose $a\omega + b\xi = (a\omega_\nu + b\xi_\nu)\varepsilon^{(\nu)} \equiv v_\nu \varepsilon^{(\nu)}$ and

$c\mathbf{W} + d\mathbf{V} = (cW^\rho + dV^\rho)\mathbf{f}_{(\rho)} \equiv U^\rho \mathbf{f}_{(\rho)}$. Then

$$\begin{aligned} e_\mu^\nu(a\omega + b\xi, c\mathbf{W} + d\mathbf{V}) &= e_\mu^\nu(v_\nu \varepsilon^{(\nu)}, U^\rho \mathbf{f}_{(\rho)}) \stackrel{(5)}{=} v_\mu U^\nu \\ &= (a\omega_\mu + b\xi_\mu)(cW^\nu + dV^\nu) = ac\omega_\mu W^\nu + ad\omega_\mu V^\nu + bc\xi_\mu W^\nu + bd\xi_\mu V^\nu \\ &= ace_\mu^\nu(\omega, \mathbf{W}) + ade_\mu^\nu(\omega, \mathbf{V}) + bce_\mu^\nu(\xi, \mathbf{W}) + bde_\mu^\nu(\xi, \mathbf{V}). \end{aligned} \quad \checkmark$$

Step 6

Define $\mathcal{B}_{\mathcal{V} \otimes \mathcal{W}^*}$ as the set of all e_μ^μ :

$$\mathcal{B}_{\mathcal{V} \otimes \mathcal{W}^*} = \left\{ e_\mu^\nu \right\}_{\mu=1}^n \left. \right|_{\nu=1}^m \quad (6)$$

In order to show that it is a basis, we must show that it spans $\mathcal{V} \otimes \mathcal{W}^*$ and is linearly independent. The proof of this involves definition (7) and formula (8).

Steps 7 – 8

$\mathcal{B}_{\mathcal{V} \otimes \mathcal{W}^*}$ spans $\mathcal{V} \otimes \mathcal{W}^*$:

For $T \in \mathcal{V} \otimes \mathcal{W}^*$, define T_ν^μ for basis pairs (7) and then extend it to covector-vectors pairs (8):

$$T_\nu^\mu \equiv T\left(\varepsilon^{(\mu)}, \mathbf{f}_{(\nu)}\right). \quad (7)$$

Let $\omega = \omega_\mu \varepsilon^{(\mu)} \in \mathcal{V}^*$ and $\mathbf{W} = W^\nu \mathbf{f}_{(\nu)} \in \mathcal{W}^*$. Then

$$\begin{aligned}
T(\omega, \mathbf{W}) &= T\left(\omega_{\mu} \varepsilon^{(\mu)}, W^{\nu} \mathbf{f}_{(\nu)}\right) \stackrel{\text{(bilinear)}}{=} \omega_{\mu} W^{\nu} T\left(\varepsilon^{(\mu)}, \mathbf{f}_{(\nu)}\right) \stackrel{(7)}{=} T_{\nu}^{\mu} \omega_{\mu} W^{\nu} \\
&\stackrel{(5)}{=} T_{\nu}^{\mu} \mathbf{e}_{\mu}^{\nu}(\omega, \mathbf{W}) \\
\Rightarrow T &= T_{\nu}^{\mu} \mathbf{e}_{\mu}^{\nu} \quad \checkmark
\end{aligned} \tag{8}$$

$\{\mathbf{e}_{\mu}^{\nu}\}$ is linearly independent:

$$x_{\rho}^{\sigma} \mathbf{e}_{\sigma}^{\rho} = 0 \quad \Rightarrow \quad x_{\rho}^{\sigma} = x_{\nu}^{\mu} \delta_{\mu}^{\sigma} \delta_{\rho}^{\nu} = x_{\nu}^{\mu} \mathbf{e}_{\mu}^{\nu}(\varepsilon^{(\sigma)}, \mathbf{f}_{(\rho)}) = 0 \quad \forall \sigma, \rho \quad \checkmark$$

Note that we have proven in passing that $\dim(\mathcal{V} \otimes \mathcal{W}^*) = nm$ since that is the size of its basis.

Step 9

$$\text{Let } \mathbf{e}_{(\mu)}^{(\nu)} = \begin{pmatrix} & \vdots & \\ \dots & 1_{\nu}^{\mu} & \dots \\ & \vdots & \end{pmatrix} \tag{9}$$

be the $n \times m$ matrix that corresponds to the linear functional \mathbf{e}_{μ}^{ν} , as constructed in Section (1) and presented in Equation (ii). There is clearly a 1-1 map between the functionals \mathbf{e}_{μ}^{ν} and the matrices $\mathbf{e}_{(\mu)}^{(\nu)}$.

Step 10

We are at last in a position to define the tensor product. Until now everything has been about linear functionals even though we have used the tensor symbol in some of the names.

Define the tensor product of a vector \mathbf{V} and covector ξ to be the matrix

$$\mathbf{V} \otimes \xi \equiv V^{\mu} \xi_{\nu} \mathbf{e}_{(\mu)}^{(\nu)}. \tag{10}$$

Recall that \mathbf{V} and ξ were defined in (3). Matrix (10) was worked out in Example 2 of Section 2:

$$\mathbf{V} \otimes \xi = \begin{pmatrix} V^1 \otimes \xi_1 & \dots & V^1 \otimes \xi_n \\ \vdots & & \vdots \\ V^n \otimes \xi_1 & \dots & V^n \otimes \xi_n \end{pmatrix}$$

Step 11

We wish to express T in (8) as linear sum of **tensors**. In (8) it is expressed as a linear sum of **linear functionals**. We do this by showing that the linear functions basis elements \mathbf{e}_μ^ν can be identified with tensor products of basis elements.

$$\text{Claim: } \mathbf{e}_{(\mu)} \otimes \varphi^\nu = \mathbf{e}_{(\mu)}^{(\nu)} \quad (11)$$

$\mathbf{e}_{(\mu)} = \delta_\mu^\alpha \mathbf{e}_{(\alpha)}$. Let $\mathbf{V} = \mathbf{e}_{(\mu)}$. Then $\mathbf{V} = V^\alpha \mathbf{e}_{(\alpha)}$. So, $V^\alpha = \delta_\mu^\alpha$. Similarly, let

$\xi = \varphi^{(\nu)}$. Then $\varphi^{(\nu)} = \delta_\beta^\nu \varphi^{(\beta)}$, $\xi = \xi_\beta \varphi^{(\beta)}$, and $\xi_\beta = \delta_\beta^\nu$. Thus,

$$\mathbf{e}_{(\mu)} \otimes \varphi^{(\nu)} = \mathbf{V} \otimes \xi = V^\alpha \xi_\beta \mathbf{e}_{(\alpha)}^{(\beta)} = \delta_\mu^\alpha \delta_\beta^\nu \mathbf{e}_{(\alpha)}^{(\beta)} = \mathbf{e}_{(\mu)}^{(\nu)}. \quad \checkmark$$

This claim may easier to visualize in matrix form, below, that references Equation (v) of Section 2:

$$\mathbf{e}_{(\mu)} \otimes \varphi^{(\nu)} \stackrel{(v)}{=} \begin{pmatrix} & \vdots & & \\ \dots & \mathbf{1}_\nu^\mu & \dots & \\ & \vdots & & \end{pmatrix} \stackrel{(9)}{=} \mathbf{e}_{(\mu)}^{(\nu)}$$

Step 12

Combining (11) with $\mathbf{T} = T_\nu^\mu \mathbf{e}_{(\mu)}^{(\nu)}$, the matrix form of (8), yields

$$\mathbf{T} = T_\nu^\mu \mathbf{e}_{(\mu)} \otimes \varphi^{(\nu)} \quad (12)$$

This expresses T in (8) as linear sum of **tensors**.

Step 13

Combining (12) with (1) allows us to express (1) in terms of tensors:

$$\mathcal{V} \otimes \mathcal{W}^* = \left\{ T_\nu^\mu \mathbf{e}_{(\mu)} \otimes \varphi^{(\nu)} \right\} \quad (13)$$

This justifies the $\mathcal{V} \otimes \mathcal{W}$ designation in (1). We have finally expressed $\mathcal{V} \otimes \mathcal{W}^*$ as a set of **tensor products**.

From (12): A $(1, 1)$ tensor is a tensor product of a vector with a covector. From (13), $\mathcal{V} \otimes \mathcal{W}^*$ is isomorphic to the n^2 dimensional vector space of tensors T_ν^μ . We know from Section 2 that the tensors T_ν^μ are $n \times n$ matrices.

Step 14

Equation (14) gives the **abstract** formula for T :

$$\begin{aligned} T(\omega, \mathbf{W}) &= T_\nu^\mu \omega_\mu W^\nu & (14) \\ T(\omega, \mathbf{W}) &= T\left(\omega_\mu \varepsilon^{(\mu)}, W^\nu \mathbf{f}_{(\nu)}\right) \stackrel{\text{(bilinearity)}}{=} \omega_\mu W^\nu T\left(\varepsilon^{(\mu)}, \mathbf{f}_{(\nu)}\right) \stackrel{(7)}{=} T_\nu^\mu \omega_\mu W^\nu. \end{aligned}$$

Steps 15 – 16

Because RHS of (14) is a sum of products, the order of the terms does not matter. But, if we were to express RHS of (14) in **matrix**-algebra terminology, the order very much matters. In what order would we express the product:

$T\omega\mathbf{W}$, $\omega\mathbf{W}T$, ωTW , or maybe $\mathbf{W}\omega T$?

Since the result must be a scalar, the order is clear. Because ω is a row vector, there must be a column vector, \mathbf{T}_ω , to its right. Because \mathbf{W} is a column vector, there must be a row vector, \mathbf{T}^W , to its left. So, it must be that

$$\mathbf{T} = \mathbf{T}_\omega \mathbf{T}^W$$

and

$$T(\omega, \mathbf{W}) = (\omega \mathbf{T}_\omega)(\mathbf{T}^W \mathbf{W}) = \omega \mathbf{T} \mathbf{W}.$$

This works in so far as $\omega \mathbf{T}_\omega$ and $\mathbf{T}^W \mathbf{W}$ are scalars. However, it doesn't work in so far as matrix multiplication is not defined between a column vector \mathbf{T}_ω and a row vector \mathbf{T}^W . In Section 2 we showed that this must be a tensor product. So, the matrix algebra version of equation (14) becomes

$$T(\omega, \mathbf{W}) = \omega \mathbf{T} \mathbf{W} \quad (15)$$

where

$$\mathbf{T} = \mathbf{T}_\omega \otimes \mathbf{T}^W. \quad (16)$$

Observe that while the positioning of ω , \mathbf{W} , and \mathbf{T} matters, the order of execution does not. One can compute $\omega \mathbf{T}_\omega$, $\mathbf{T}_\omega \otimes \mathbf{T}^W$, and $\mathbf{T}^W \mathbf{W}$ in any order.

An Aside

Although a tensor T has been defined as a linear function acting on a pair (ω, \mathbf{W}) , there is nothing forcing it to act on both. For example, we can think of T as a linear function from \mathcal{W} to \mathcal{W} as follows. Since $\mathbf{T}^W \mathbf{W}$ is a scalar, say K , then

$\mathbf{T}\mathbf{W} = \mathbf{T}_\omega \otimes \mathbf{T}^W \mathbf{W} = \mathbf{T}_\omega \otimes K = K \mathbf{T}_\omega = (\mathbf{T}^W \mathbf{W}) \mathbf{T}_\omega$ is a column vector. So

$T|_{\mathcal{W}}: \mathcal{W} \rightarrow \mathcal{W}: T(\mathbf{W}) = \mathbf{T}\mathbf{W}$. Similarly we can think of T as a linear function from \mathcal{V}^* to \mathcal{V}^* .

One more remark about T . Note that while the linear functional version of T acts on the product of a covector with a vector, the matrix version of \mathbf{T} is itself the opposite, a product of a vector with a covector.

Step 17

Not only can a tensor act on vectors and covectors, a tensor can also act on another tensor to produce a tensor, even if the tensors have different domains.

For example, suppose $\mathbf{R} = R_\nu^\mu \mathbf{e}_{(\mu)} \otimes \varphi^{(\nu)} \in \mathcal{V} \otimes \mathcal{W}^*$ and $\mathbf{S} = S_\rho^\sigma \mathbf{f}_{(\sigma)} \otimes \varepsilon^{(\rho)} \in \mathcal{W} \otimes \mathcal{V}^*$ are tensors. Define a $(2, 2)$ tensor in $\mathcal{V} \otimes \mathcal{W} \otimes \mathcal{W}^* \otimes \mathcal{V}^*$ by

$$\begin{aligned} \mathbf{T} &= \mathbf{R} \otimes \mathbf{S} = R_\nu^\mu \mathbf{e}_{(\mu)} \otimes \varphi^{(\nu)} \otimes S_\rho^\sigma \mathbf{f}_{(\sigma)} \otimes \varepsilon^{(\rho)} \\ &\equiv R_\nu^\mu S_\rho^\sigma \mathbf{e}_{(\mu)} \otimes \mathbf{f}_{(\sigma)} \otimes \varphi^{(\nu)} \otimes \varepsilon^{(\rho)} = T_{\nu\rho}^{\mu\sigma} \mathbf{e}_{(\mu)} \otimes \mathbf{f}_{(\sigma)} \otimes \varphi^{(\nu)} \otimes \varepsilon^{(\rho)} \end{aligned} \quad (17)$$

where $T_{\nu\rho}^{\mu\sigma} = R_\nu^\mu S_\rho^\sigma$.

We can view this as the tensor \mathbf{R} acting on the tensor \mathbf{S} to produce the tensor \mathbf{T} .

The current formula is a special case of the product of two tensors, a gentle introduction to the more general formula (17) in Section 6. The equation (17) above can be used, for example, to compute $\mathbf{R} \otimes \mathbf{S}(\omega, \xi, \mathbf{W}, \mathbf{V})$:

$$\mathbf{R} \otimes \mathbf{S}(\omega, \xi, \mathbf{W}, \mathbf{V}) = \mathbf{T}(\omega, \xi, \mathbf{W}, \mathbf{V}) \stackrel{(14)}{=} T_{\nu\rho}^{\mu\sigma} \omega_\mu \xi_\sigma W^\nu V^\rho \stackrel{(17)}{=} R_\nu^\mu S_\rho^\sigma \omega_\mu \xi_\sigma W^\nu V^\rho.$$

5 Lower Rank Tensors

Before we move to the general case of a (k, ℓ) tensor, we quickly summarize the lower tensors.

A $(2, 0)$ tensor space is $\mathcal{V} \otimes \mathcal{W} = \{T : \mathcal{V}^* \times \mathcal{W}^* \rightarrow \mathbf{F} : T \text{ is bilinear}\}$.

A $(2, 0)$ tensor is formed from a tensor product of an n vector with an m vector. The result is that a $(2, 0)$ tensor is an nm vector. As noted earlier, some books may treat a $(2, 0)$ tensor as an $n \times m$ matrix.

A $(0, 2)$ tensor space is $\mathcal{V}^* \otimes \mathcal{W}^* = \{T : \mathcal{V} \times \mathcal{W} \rightarrow \mathbf{F} : T \text{ is bilinear}\}$.

A $(0, 2)$ tensor is an nm covector.

A $(1, 0)$ tensor space is $\mathcal{V} = \{T : \mathcal{V}^* \rightarrow \mathbf{F} : T \text{ is linear}\}$.

A $(1, 0)$ tensor is a vector because there is no product involved.

A $(0, 1)$ tensor space is $\mathcal{V}^* = \{T : \mathcal{V} \rightarrow \mathbf{F} : T \text{ is linear}\}$

A $(0, 1)$ tensor is a covector.

For a $(0, 0)$ tensor, there is no functional definition at all.

A $(0, 0)$ tensor is defined to be a scalar.

6 Rank (k, ℓ) Tensors

We can develop the equations of (k, ℓ) tensors with each vector space \mathcal{V}_i and covector space \mathcal{W}_j distinct, but the notation quickly explodes to unreadability. Therefore, we now assume that all of the vector and covector spaces are copies from a single vector space \mathcal{V} of dimension n ; i.e., $\mathcal{V}_i = \mathcal{W}_j = \mathcal{V}$ for all i, j .

We develop rank (k, ℓ) tensors following the steps (1–17) used for rank $(1, 1)$. Since all of the proofs are straight-forward extensions of the proofs given in Section 4, they are omitted. We begin with equation (1), definition of a tensor.

$$\begin{aligned} & \mathcal{V}_1 \otimes \cdots \otimes \mathcal{V}_k \otimes \mathcal{W}_1^* \otimes \cdots \otimes \mathcal{W}_\ell^* \\ & \equiv \{T : \mathcal{V}_1^* \times \cdots \times \mathcal{V}_k^* \times \mathcal{W}_1 \times \cdots \times \mathcal{W}_\ell \rightarrow \mathbf{F} : T \text{ is multilinear}\}. \end{aligned} \quad (1)$$

To simplify this, we set

$$\mathcal{V} = \mathcal{V}_1 \times \cdots \times \mathcal{V}_k,$$

$$\mathcal{W} = \mathcal{W}_1 \times \cdots \times \mathcal{W}_\ell,$$

and

$$T' = \mathcal{V}_1 \otimes \cdots \otimes \mathcal{V}_k \otimes \mathcal{W}_1^* \otimes \cdots \otimes \mathcal{W}_\ell^*$$

This enables us to rewrite (1) to more closely resemble how we expressed (1) for $(1, 1)$ tensors:

$$T' \equiv \{\mathcal{T} : \mathcal{V}^* \times \mathcal{W} \rightarrow \mathbf{F} : T \text{ is multilinear}\} \quad (1')$$

Since there is only one reference vector space \mathcal{V} , equations (2) and (3) can be expressed simply. We specify just a single basis for all the \mathcal{V}_i and a single basis for all the \mathcal{W}_j :

$$\mathcal{B} = \left\{ \mathbf{e}_{(\mu)} \right\}_{\mu=1}^n \text{ and } \mathcal{B}^* = \left\{ \boldsymbol{\varepsilon}^{(\nu)} \right\}_{\nu=1}^n \quad (2)$$

$$\begin{aligned} \mathbf{V} &= V^\mu \mathbf{e}_{(\mu)} \in \mathcal{V}, \quad \mathbf{W} = W^\rho \mathbf{e}_{(\rho)} \in \mathcal{W}, \\ \boldsymbol{\omega} &= \omega_\nu \boldsymbol{\varepsilon}^{(\nu)} \in \mathcal{V}^*, \text{ and } \boldsymbol{\xi} = \xi_\sigma \boldsymbol{\varepsilon}^{(\sigma)} \in \mathcal{W}^* \end{aligned} \quad (3)$$

We define the linear functionals on the basis elements in (4) and then extend them to all members of $\mathcal{V}^* \times \mathcal{W}$ in (5):

$$\mathbf{e}_{\mu_1 \cdots \mu_k}^{\nu_1 \cdots \nu_\ell} \left(\boldsymbol{\varepsilon}^{(\sigma_1)}, \dots, \boldsymbol{\varepsilon}^{(\sigma_k)}, \mathbf{e}_{(\rho_1)}, \dots, \mathbf{e}_{(\rho_\ell)} \right) \equiv \delta_{\mu_1}^{\sigma_1} \cdots \delta_{\mu_k}^{\sigma_k} \delta_{\rho_1}^{\nu_1} \cdots \delta_{\rho_\ell}^{\nu_\ell}. \quad (4)$$

$$\mathbf{e}_{\mu_1 \cdots \mu_k}^{\nu_1 \cdots \nu_\ell} : \mathcal{V}^* \times \mathcal{W} \rightarrow \mathbf{F} : \mathbf{e}_{\mu_1 \cdots \mu_k}^{\nu_1 \cdots \nu_\ell} \left(\boldsymbol{\omega}^{(1)}, \dots, \boldsymbol{\omega}^{(k)}, \mathbf{W}_{(1)}, \dots, \mathbf{W}_{(\ell)} \right) = \omega_{\mu_1} \cdots \omega_{\mu_k} W^{\nu_1} \cdots W^{\nu_\ell} \quad (5)$$

We define the basis for \mathcal{T}' and expressions for the components.

$$\mathcal{B}_{\mathcal{T}'} = \left\{ \mathbf{e}_{\mu_1 \cdots \mu_k}^{\nu_1 \cdots \nu_\ell} : \mu_1 = 1, \dots, n; \dots; \mu_k = 1, \dots, n; \nu_1 = 1, \dots, n; \dots; \nu_\ell = 1, \dots, n \right\} \quad (6)$$

$$T_{\nu_1 \cdots \nu_\ell}^{\mu_1 \cdots \mu_k} = T\left(\varepsilon^{(\mu_1)}, \dots, \varepsilon^{(\mu_k)}, \mathbf{e}_{(\nu_1)}, \dots, \mathbf{e}_{(\nu_\ell)}\right). \quad (7)$$

$$T = T_{\nu_1 \cdots \nu_\ell}^{\mu_1 \cdots \mu_k} \mathbf{e}_{\mu_1 \cdots \mu_k}^{\nu_1 \cdots \nu_\ell} \quad (8)$$

Since the basis has size $n^{k+\ell}$, $\dim(\mathcal{T}') = n^{k+\ell}$.

There is a 1-1 map between the **functionals** $\mathbf{e}_{\mu_1 \cdots \mu_k}^{\nu_1 \cdots \nu_\ell}$ and the **matrices** $\mathbf{e}_{(\mu_1) \cdots (\mu_k)}^{(\nu_1) \cdots (\nu_\ell)}$ (9)

Each $\mathbf{e}_{(\mu_1) \cdots (\mu_k)}^{(\nu_1) \cdots (\nu_\ell)}$ is an $n^k \times n^\ell$ matrix having all zeros except for a "1" in the cell

(n^k, n^ℓ) . There are $n^{k+\ell}$ basis matrices $\mathbf{e}_{(\mu_1) \cdots (\mu_k)}^{(\nu_1) \cdots (\nu_\ell)}$. The parentheses on the indices indicate that we are enumerating matrices and not matrix components.

Define the tensor product of basis vectors and basis covectors:

$$\mathbf{e}_{(\mu_1)} \otimes \cdots \otimes \mathbf{e}_{(\mu_k)} \otimes \varepsilon^{(\nu_1)} \otimes \cdots \otimes \varepsilon^{(\nu_\ell)} = \mathbf{e}_{(\mu_1) \cdots (\mu_k)}^{(\nu_1) \cdots (\nu_\ell)}: \quad (10)$$

As shown for (1, 1) tensors, this leads to the tensor product of vectors and covectors:

$$\mathbf{V}_{(1)} \otimes \cdots \otimes \mathbf{V}_{(k)} \otimes \xi^{(1)} \otimes \cdots \otimes \xi^{(\ell)} \equiv V^{\mu_1} \cdots V^{\mu_k} \xi_{\nu_1} \cdots \xi_{\nu_\ell} \mathbf{e}_{(\mu_1) \cdots (\mu_k)}^{(\nu_1) \cdots (\nu_\ell)}. \quad (11)$$

Substituting (10) into (8) yields

$$T = T_{\nu_1 \cdots \nu_\ell}^{\mu_1 \cdots \mu_k} \mathbf{e}_{(\mu_1)} \otimes \cdots \otimes \mathbf{e}_{(\mu_k)} \otimes \varepsilon^{(\nu_1)} \otimes \cdots \otimes \varepsilon^{(\nu_\ell)} \quad (12)$$

and substituting (12) into (1') yields

$$\mathcal{T}' = \left\{ T_{\nu_1 \cdots \nu_\ell}^{\mu_1 \cdots \mu_k} \mathbf{e}_{(\mu_1)} \otimes \cdots \otimes \mathbf{e}_{(\mu_k)} \otimes \varepsilon^{(\nu_1)} \otimes \cdots \otimes \varepsilon^{(\nu_\ell)} \right\}. \quad (13)$$

From (12), a (k, ℓ) tensor is a tensor product of k vectors and ℓ covectors. It is a $n^k \times n^\ell$ matrix whose elements are the components $T_{\nu_1 \cdots \nu_\ell}^{\mu_1 \cdots \mu_k}$.

From (13), $\mathcal{T}' = \mathbf{V}_1 \otimes \cdots \otimes \mathbf{V}_k \otimes \mathbf{W}_1^* \otimes \cdots \otimes \mathbf{W}_\ell^*$ is isomorphic to the $n^{(k+\ell)}$ -dimensional vector space of matrices $T = T_{\nu_1 \cdots \nu_\ell}^{\mu_1 \cdots \mu_k} \mathbf{e}_{(\mu_1)} \otimes \cdots \otimes \mathbf{e}_{(\mu_k)} \otimes \varepsilon^{(\nu_1)} \otimes \cdots \otimes \varepsilon^{(\nu_\ell)}$.

Since the basis vectors are composed of all zeroes except for a single "1", all of

the information is contained in the components $T_{\nu_1 \dots \nu_k}^{\mu_1 \dots \mu_\ell}$. Thus, we loosely say that

\mathbf{T}' is isomorphic to the vector space of $(n^k \times n^\ell)$ matrices $T_{\nu_1 \dots \nu_k}^{\mu_1 \dots \mu_\ell}$.

As with $(1, 1)$ tensors, we also have the formula

$$T(\omega^{(1)}, \dots, \omega^{(k)}, \mathbf{W}_{(1)}, \dots, \mathbf{W}_{(\ell)}) = T_{\nu_1 \dots \nu_\ell}^{\mu_1 \dots \mu_k} \omega_{\mu_1} \dots \omega_{\mu_k} W^{\nu_1} \dots W^{\nu_\ell} \quad (14)$$

and its matrix form (on RHS)

$$T(\omega^{(1)}, \dots, \omega^{(k)}, \mathbf{W}_{(1)}, \dots, \mathbf{W}_{(\ell)}) = \omega^{(1)} \otimes \dots \otimes \omega^{(k)} \mathbf{T} \mathbf{W}_{(1)} \otimes \dots \otimes \mathbf{W}_{(\ell)}. \quad (15)$$

where \mathbf{T} is a $n^k \times n^\ell$ matrix.

Also, \mathbf{T} can be expressed as the tensor product of k vectors $\mathbf{T}_{(i)}$ and ℓ covectors $\mathbf{T}^{(j)}$:

$$\mathbf{T} = \mathbf{T}_{(k)} \otimes \dots \otimes \mathbf{T}_{(1)} \otimes \mathbf{T}^{(\ell)} \otimes \dots \otimes \mathbf{T}^{(1)}. \quad (16)$$

This makes sense because $\omega = \omega^{(1)} \otimes \dots \otimes \omega^{(k)}$ is a length n^k row vector, \mathbf{T} is a $n^k \times n^\ell$ matrix, and $\mathbf{W} = \mathbf{W}_{(1)} \otimes \dots \otimes \mathbf{W}_{(\ell)}$ is length n^ℓ column vector. So it is legitimate to perform the matrix operations $\omega \mathbf{T} \mathbf{W}$ even though matrices ω , \mathbf{T} , and \mathbf{W} are themselves tensor products. The indices on \mathbf{T} in (16) are in reverse numerical order because $\omega^{(k)} \mathbf{T}_{(k)}$ is performed 1st, reducing to a scalar, then $\omega^{(k-1)} \mathbf{T}_{(k-1)}$ reduces to a scalar, etc., and similarly $\mathbf{T}^{(1)} \mathbf{W}_{(1)}$ becomes a scalar, then $\mathbf{T}^{(2)} \mathbf{W}_{(2)}$, ..., and last, $\mathbf{T}^{(\ell)} \mathbf{W}_{(\ell)}$. The various matrix and tensor products can be performed in any order.

Just because a tensor T has been defined on k covectors and ℓ vectors does not mean it cannot act on a subset of covectors and/or vectors. For example,

$$T_{\nu_1 \dots \nu_\ell}^{\mu_1 \dots \mu_k} \omega_{\mu_2} \dots \omega_{\mu_k} W^{\nu_2} \dots W^{\nu_\ell} = (\text{scalar}) T_{\nu_1}^{\mu_1} \text{ is a } (1, 1) \text{ tensor.}$$

Moreover, not only can a tensor act on covectors and vectors, but any tensor can act on any other tensor to produce a tensor as follows.

Definition. If $\mathbf{R} = R_{\nu_1 \dots \nu_\ell}^{\mu_1 \dots \mu_k} \mathbf{e}_{(\mu_1)} \otimes \dots \otimes \mathbf{e}_{(\mu_k)} \otimes \varepsilon^{(\nu_1)} \otimes \dots \otimes \varepsilon^{(\nu_\ell)}$ is a (k, ℓ) tensor and

$\mathbf{S} = S_{\nu_{\ell+1} \dots \nu_{\ell+n}}^{\mu_{k+1} \dots \mu_{k+m}} \mathbf{e}_{(\mu_{k+1})} \otimes \dots \otimes \mathbf{e}_{(\mu_{k+m})} \otimes \varepsilon^{(\nu_{\ell+1})} \otimes \dots \otimes \varepsilon^{(\nu_{\ell+n})}$ is an (m, n) tensor, define a

$(k+m, \ell+n)$ tensor $\mathbf{T} = \mathbf{R} \otimes \mathbf{S}$ as

$$\begin{aligned}
\mathbf{T} &= \mathbf{R} \otimes \mathbf{S} = R_{\nu_1 \cdots \nu_\ell}^{\mu_1 \cdots \mu_k} \mathbf{e}_{(\mu_1)} \otimes \cdots \otimes \mathbf{e}_{(\mu_k)} \otimes \varepsilon^{(\nu_1)} \otimes \cdots \otimes \varepsilon^{(\nu_\ell)} \\
&\quad \otimes S_{\nu_{\ell+1} \cdots \nu_{\ell+n}}^{\mu_{k+1} \cdots \mu_{k+m}} \mathbf{e}_{(\mu_{k+1})} \otimes \cdots \otimes \mathbf{e}_{(\mu_{k+m})} \otimes \varepsilon^{(\nu_{\ell+1})} \otimes \cdots \otimes \varepsilon^{(\nu_{\ell+n})} \\
&\equiv R_{\nu_1 \cdots \nu_\ell}^{\mu_1 \cdots \mu_k} S_{\nu_{\ell+1} \cdots \nu_{\ell+n}}^{\mu_{k+1} \cdots \mu_{k+m}} \mathbf{e}_{(\mu_1)} \otimes \cdots \otimes \mathbf{e}_{(\mu_{k+m})} \otimes \varepsilon^{(\nu_1)} \otimes \cdots \otimes \varepsilon^{(\nu_{\ell+n})} \\
&= T_{\nu_1 \cdots \nu_\ell}^{\mu_1 \cdots \mu_k} \mathbf{e}_{(\mu_1)} \otimes \cdots \otimes \mathbf{e}_{(\mu_{k+m})} \otimes \varepsilon^{(\nu_1)} \otimes \cdots \otimes \varepsilon^{(\nu_{\ell+n})}
\end{aligned} \tag{17}$$

where $T_{\nu_1 \cdots \nu_\ell}^{\mu_1 \cdots \mu_k} = R_{\nu_1 \cdots \nu_\ell}^{\mu_1 \cdots \mu_k} S_{\nu_{\ell+1} \cdots \nu_{\ell+n}}^{\mu_{k+1} \cdots \mu_{k+m}}$.

Then,

$$\begin{aligned}
\mathbf{R} \otimes \mathbf{S} \left(\omega^{(1)}, \dots, \omega^{(k+\ell)}, \mathbf{W}_{(1)}, \dots, \mathbf{W}_{(m+n)} \right) &= \mathbf{T} \left(\omega^{(1)}, \dots, \omega^{(k+\ell)}, \mathbf{W}_{(1)}, \dots, \mathbf{W}_{(m+n)} \right) \\
&\stackrel{(14)}{=} T_{\nu_1 \cdots \nu_{m+n}}^{\mu_1 \cdots \mu_{k+\ell}} \omega_{\mu_1} \cdots \omega_{\mu_{k+\ell}} W^{\nu_1} \cdots W^{\nu_{m+n}} = R_{\nu_1 \cdots \nu_\ell}^{\mu_1 \cdots \mu_k} S_{\nu_{\ell+1} \cdots \nu_{\ell+n}}^{\mu_{k+1} \cdots \mu_{k+m}} \omega_{\mu_1} \cdots \omega_{\mu_{k+\ell}} W^{\nu_1} \cdots W^{\nu_{m+n}}.
\end{aligned} \tag{18}$$

This can also be viewed as tensor \mathbf{R} acting on tensor \mathbf{S} to produce tensor \mathbf{T} , or as tensor \mathbf{S} acting on tensor \mathbf{R} to produce tensor \mathbf{T} .

Note that all of the covector are grouped together as are the vectors in both (17) and (18).

9 Examples

Example 1. A (2,2) tensor is a quad-linear map

$$T : \mathcal{V}^* \times \mathcal{V}^* \times \mathcal{V} \times \mathcal{V} \rightarrow \mathbf{F} : T(\omega, \xi, V, W) = {}_{\mu_1}^{\omega} \xi_{\mu_2} {}^{(5)}_{\nu_1 \nu_2} T^{\mu_1 \mu_2} V^{\nu_1} W^{\nu_2}.$$

To neutralize covectors ω and ξ , they must be multiplied on their right by two vectors. To neutralize vectors V and W , they must be multiplied on their left by two row vectors. Thus, a (2,2) tensor T is the tensor product of the tensor product of two vectors with the tensor product of two covectors:

$$T(\omega, \xi, V, W) = \omega \otimes \xi T V \otimes W$$

where

$$T = T_{(\xi)} \otimes T_{(\omega)} \otimes T^{(W)} \otimes T^{(V)}.$$

Example 2. Quantum Mechanics. $\mathbf{F} = \mathbb{C}$.

Alice and Bob each measure the spin state of an electron. Let S_A represent Alice's state space, the vector space of all her states. Let S_B represent Bob's state space. Let $|A\rangle$ and $|B\rangle$ be state vectors in S_A and S_B , respectively, and let $\{|a\rangle\}$ and $\{|b\rangle\}$ be bases for S_A and S_B . Then $S_{AB} = S_A \otimes S_B$ is the vector space whose basis is the set of tensor product states $|ab\rangle \equiv |a\rangle \otimes |b\rangle$. These basis objects are vectors since they are tensor products of vectors. Similarly,

if $|A\rangle = \sum_a \alpha_a |a\rangle$ and $|B\rangle = \sum_b \beta_b |b\rangle$, then

$$|AB\rangle = |A\rangle \otimes |B\rangle = \sum_a \sum_b \alpha_a \beta_b |ab\rangle \in S_{AB}$$

are vectors.

However, S_{AB} is larger than the set of vectors $\{|AB\rangle\}$. The singlet state is a vector of the form $|\Psi\rangle = \sum_a \sum_b \psi_{ab} |ab\rangle \in S_{AB}$ that cannot be expressed as $|A\rangle \otimes |B\rangle$. We give a more detailed example of this behavior next.

Example 3. General Relativity. $\mathbf{F} = \mathbb{R}$.

Let $e_{(0)} = \begin{pmatrix} i \\ 0 \\ 0 \\ 0 \end{pmatrix}$, $e_{(1)} = \begin{pmatrix} 0 \\ 1 \\ 0 \\ 0 \end{pmatrix}$, $e_{(2)} = \begin{pmatrix} 0 \\ 0 \\ 1 \\ 0 \end{pmatrix}$, and $e_{(3)} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix}$ be a basis for

spacetime, S . Let $\varepsilon^{(\nu)} = e_{(\nu)}^T$. For example, $\varepsilon^{(0)} = \begin{pmatrix} i & 0 & 0 & 0 \end{pmatrix}$.

Then $\left\{ \mathbf{e}_{(\mu)} \otimes \varepsilon^{(\nu)} \right\}_{\mu, \nu=0}^3$ is a basis for $S^2 = S \otimes S$, where elements of the basis are of the form

$$\mathbf{e}_{(\mu)}^{(\nu)} \equiv \mathbf{e}_{(\mu)} \otimes \varepsilon^{(\nu)} = \begin{pmatrix} 0 & \vdots & 0 \\ \cdots & \mathbf{e}^\mu \varepsilon_\nu & \cdots \\ 0 & \vdots & 0 \end{pmatrix}.$$

Define the metric tensor

$$\eta = (\eta^\mu_\nu) = \begin{pmatrix} -1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}.$$

$\eta = \sum_{\mu} \sum_{\nu} \eta^\mu_\nu \mathbf{e}_{(\mu)} \otimes \varepsilon^{(\nu)}$ is a member of S^2 that cannot be expressed as a product tensor. That is, let $\alpha = \sum_{\mu} \alpha^\mu \mathbf{e}_{(\mu)}$ and $\beta = \sum_{\nu} \beta_\nu \varepsilon^{(\nu)}$. Then $\eta \neq \alpha \otimes \beta$:

Suppose

$$\eta = \alpha \otimes \beta = \sum_{\mu} \sum_{\nu} \alpha^\mu \beta_\nu \mathbf{e}_{(\mu)} \otimes \varepsilon^{(\nu)} = \begin{pmatrix} \alpha^0 \beta_0 & \cdots & \alpha^0 \beta_3 \\ \vdots & & \vdots \\ \alpha^3 \beta_0 & \cdots & \alpha^3 \beta_3 \end{pmatrix}.$$

Then

$$-1 = \alpha^0 \beta_0 \Rightarrow \alpha^0 \neq 0 \text{ and } 1 = \alpha^1 \beta_1 \Rightarrow \beta_1 \neq 0.$$

Yet,

$$\alpha^0 \beta_1 = 0,$$

a contradiction.

So the space S^2 generated by the basis $\left\{ \mathbf{e}_{(\mu)} \otimes \varepsilon^{(\nu)} \right\}_{\mu, \nu=0}^3$ is larger than the space of product vectors $\{\alpha \otimes \beta\}$. Using quantum mechanics terminology we would say that the metric tensor is **entangled**.

Also, as mentioned, some books may label the metric tensor as the matrix $\eta_{\mu\nu}$. I prefer to write $\eta_{\mu\nu}$ as a (1×16) covector:

$$\begin{pmatrix} -1 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 1 \end{pmatrix}.$$

I would express $\eta^{\mu\nu}$ as a (16×1) column vector, and then the product $\eta_{\mu\nu} \eta^{\mu\nu}$ would be a scalar, not another matrix as would be the case if both tensors were matrices.

Finally, the metric tensor enables compact notation for things like the space-time interval between two events:

$$s^2 = -(\mathbf{c}\Delta t)^2 + (\Delta x)^2 + (\Delta y)^2 + (\Delta z)^2.$$

To write this, we let $\Delta \mathbf{x} \equiv \begin{pmatrix} \mathbf{c}\Delta t \\ \Delta x \\ \Delta y \\ \Delta z \end{pmatrix}$. Then $s^2 = (\Delta \mathbf{x})^\top \eta (\Delta \mathbf{x}) = \eta_\nu^\mu (\Delta x)_\mu^\top (\Delta x)^\nu$.