

1 What is a tensor?

Let us define a tensor of rank $r \in \mathbb{N}$ to be the data:

- A set of numbers $s_1, s_2, \dots, s_r \in \mathbb{N}$, where s_i is said to be the size of the tensor along dimension i .
- A function $F : [s_1] \times [s_2] \times \dots \times [s_r] \rightarrow (\mathbb{R} \rightarrow \mathbb{R})$, where $[u] \equiv \{1, 2, \dots, u\}$.

Given a tensor $T \equiv (r, S \equiv (s_1, s_2, \dots, s_r), F)$, we denote by:

$$\begin{aligned} T[ix_1, ix_2, \dots, ix_r] &: ([s_1] \times [s_2] \times \dots \times [s_r]) \rightarrow (\mathbb{R} \rightarrow \mathbb{R}) \\ T[ix_1, ix_2, \dots, ix_r] &\equiv F(ix_1, ix_2, ix_3, \dots, ix_r) \end{aligned}$$

Let us now instantiate an honest to god tensor. We shall create what plebes know as a "vector field" to be a tensor.

1.1 Example: Vector field as a tensor

Consider the vector field on \mathbb{R}^2 to be $V \equiv (\sin(x), \cos(y))$.

For us, this will correspond to a tensor of rank 1 $T \equiv (r \equiv 1, S \equiv (2), F)$ where:

$$F(1) \equiv \lambda x. \sin x \quad F(2) \equiv \lambda x. \cos x$$

1.2 Example: Function as a tensor

A function $f : \mathbb{R} \rightarrow \mathbb{R}$ is a tensor of rank 1 $T_f \equiv (r \equiv 1, S \equiv (1), F(1) \equiv f)$.

1.3 Example: Scalar as a tensor

A scalar $r \in \mathbb{R}$ is a tensor of rank 1 $T \equiv (r \equiv 0, S \equiv (), F() \equiv \lambda_.r)$

1.4 Matrix field as a tensor

The matrix field which maps each point (x, y) to the matrix $\begin{bmatrix} x & 0 \\ 0 & y \end{bmatrix}$ is a rank 2 tensor:

$$\begin{aligned} T &\equiv (r \equiv 2, S \equiv (2, 2), F) \\ F(1, 1) &\equiv \lambda x.x \quad F(1, 2) \equiv \lambda_.0 \\ F(2, 1) &\equiv \lambda_.0 \quad F(2, 2) \equiv \lambda y.y \end{aligned}$$

2 Tensor derivatives

We are often interested in understanding how one tensor varies with respect to another. But what does this question even mean? Well, I claim there is only one sensible explanation. A tensor after all is just a collection of functions. So the derivative of one tensor with respect to another, say $\frac{\partial A[i_1, i_2, \dots, i_n]}{\partial B[j_1, j_2, \dots, j_m]}$ can only be a new tensor whose entries are the derivatives of *each* function in A with *each* function in B .

This instantly leads to the definition:

$$\begin{aligned} A &\equiv (n, S_A, F_A) & B &\equiv (m, S_B, F_b) \\ C &\equiv (n + m, (S_A, S_B), F) \\ F(i_1, i_2, \dots, i_n, j_1, j_2, \dots, j_m) &\equiv \frac{\partial A[i_1, i_2, \dots, i_n]}{\partial B[j_1, j_2, \dots, j_m]} \end{aligned}$$

Note that this is perfectly well defined, since $A[i_1, i_2, \dots, i_n] : \mathbb{R} \rightarrow \mathbb{R}$. Similarly, $B[j_1, j_2, \dots, j_m] : \mathbb{R} \rightarrow \mathbb{R}$, and we hopefully know how to differentiate single variable functions.

The C as written above is often colourfully written as:

$$\frac{\partial A}{\partial B} \frac{\partial A[i_1, i_2, \dots, i_n]}{\partial B[j_1, j_2, \dots, j_m]}$$

and many other abuses of notation. But never forget what it is doing: It is simply creating a convenient way to consider the change of every component of A relative to every component of B .

3 The derivative $\frac{\partial x^T x}{\partial x}$

Notice that if x is an honest to god vector, the above expression makes no sense. For example, let $x = (5, 5) \in \mathbb{R}^2$. Now, $x^T x = 50$, leading to the absurd expression $\frac{\partial 50}{\partial (5, 5)}$, which is quite senseless since differentiation is only defined for *functions*.

Hence, whenever people write such expressions, they really mean a rank 1, shape (n) tensor. that is, an n -tuple of scalar functions, each function describing the value of the n th component of the vector, relative to some parametrization.

Let $x \equiv (r \equiv 1, S \equiv (2), F_x)$. Now, $x^T x \equiv (r \equiv 0, S \equiv ()), F_{xtx} \equiv F_x(1)F_x(1) + F_x(2)F_x(2) = F_x(1)^2 + F_x(2)^2$

Let us now calculate $\frac{\partial x^T x}{\partial x}$ as we have agreed upon above:

$$\begin{aligned}
& \frac{\partial x^T x}{\partial x}(r \equiv 0 + 1, S \equiv 2, F_{der}) \\
F_{der} \llbracket 1 \rrbracket &= \frac{\partial F_{xtx}}{\partial F_x(1)} = \frac{\partial F_x(1)^2 + F_x(2)^2}{\partial F_x(1)} = 2F_x(1) \\
F_{der} \llbracket 2 \rrbracket &= \frac{\partial F_{xtx}}{\partial F_x(2)} = \frac{\partial F_x(1)^2 + F_x(2)^2}{\partial F_x(2)} = 2F_x(2)
\end{aligned}$$

Hence, $\frac{\partial x^T x}{\partial x} = 2 \cdot x$.

We can show that the collection of tensors form a vector space over \mathbb{R} , since functions $\mathbb{R} \rightarrow \mathbb{R}$ form a vector space over \mathbb{R} , and a tensor is a clever collection of such scalar functions. Hence, the notation $2 \cdot x$ is interpreted in terms of this vector space structure.