Lecture 18: Grad, Div, Curl, and all that...

# Grad, Div, Curl, and all that<sup>1</sup>⁄<sub>4</sub>

Reading:Mathews & Walker, Mathematical Methods of Physics, ch. 15.S. Weinberg, Gravitation and Cosmology, ch. 3 & 4.Ohanian and Ruffini, Gravitation and Spacetime, ch. 6 & 7.McConnell, Applications of Tensor Analysis, Ch. 12.

#### The generally covariant differential operators

The gradient operator is obvious, and we have already derived it. If  $\phi(x)$  is a scalar field, then

$$\varphi_{,\mu} \stackrel{\text{\tiny def}}{=} \frac{\partial \varphi}{\partial x^{\mu}} \tag{18.1}$$

in fact transforms as a covariant vector under general coordinate transformations.

Next, consider the covariant curl defined by

$$\operatorname{curl}_{\nu} A_{\mu} \stackrel{ar}{=} A_{\mu;\nu} - A_{\nu;\mu} = A_{\mu,\nu} - A_{\nu,\mu} - \left[ \begin{cases} \lambda \\ \mu\nu \end{cases} - \begin{cases} \lambda \\ \nu\mu \end{cases} \right] A_{\lambda}$$
$$\equiv A_{\mu,\nu} - A_{\nu,\mu} . \tag{18.2}$$

Finally, the divergence should generalize the flat-space result div  $V = V^{\mu}_{,\mu}$ :

$$\operatorname{div} V \stackrel{df}{=} V^{\mu}_{;\,\mu} \equiv V^{\mu}_{,\,\mu} + \begin{cases} \mu \\ \mu\sigma \end{cases} V^{\sigma} \,. \tag{18.2}$$

However,

$$\left\{ \begin{array}{l} \mu \\ \mu\sigma \end{array} \right\} = \frac{1}{2} g^{\mu\lambda} \left[ g_{\mu\lambda,\sigma} + g_{\lambda\sigma,\mu} - g_{\mu\sigma,\lambda} \right] \equiv \frac{1}{2} g^{\mu\lambda} g_{\mu\lambda,\sigma} \tag{18.3}$$

where we have used the antisymmetry in  $\mu\lambda$  of the terms  $g_{\lambda\sigma,\mu} - g_{\mu\sigma,\lambda}$  to drop them after contraction with  $g^{\mu\lambda}$ .

Now consider  $g_{\mu\nu}$  as a matrix G, with  $dG = g_{\mu\nu,\sigma} dx^{\sigma}$ ; then

$$\begin{cases} \mu \\ \mu\sigma \end{cases} dx^{\sigma} = \frac{1}{2} \operatorname{Trace} \left[ G^{-1} dG \right]. \tag{18.4}$$

Let us call det(*G*) = -g (this is a standard notation, with the  $\sim$  sign introduced to make *g* positive); then

$$-g = e^{\operatorname{Irace}\left[\log(G)\right]} \tag{18.5}$$

The divergence theorem

(Eq. 18.5 is far from obvious, useful, and *worth remembering*. The proof is given below. For now, just believe it!)

Thus

$$\log(-g) = \operatorname{Tr}\left[\log(G)\right]$$
  
and  
$$\operatorname{Tr}\left[\log(G+dG)\right] = \operatorname{Tr}\left[\log(G)\right] + \operatorname{Tr}\left[\log(1+G^{-1}dG)\right],$$
  
hence  
$$\log\left(-g-dg\right) = \log\left(-g\right) + \operatorname{Tr}\left[G^{-1}dG\right]$$
  
so that  
$$\left\{ \begin{array}{c} \mu\\ \mu\sigma \end{array} \right\} dx^{\sigma} = \frac{1}{2} d\left[\log(-g)\right] \equiv d\left[\log(\sqrt{-g})\right]$$
  
or (we can now drop the ~ sign from  $\sqrt{-g}$ )  
$$\left\{ \begin{array}{c} \mu\\ \mu\sigma \end{array} \right\} = \partial_{\sigma} \log(\sqrt{g}) = \frac{1}{\sqrt{g}} \partial_{\sigma} \sqrt{g} .$$
 (18.6)

"So what?" you may say, "I've got my own troubles." Here's what:

Remember Eq. 18.2? Now we may write

$$V^{\mu}_{;\,\mu} = V^{\mu}_{,\,\mu} + \frac{1}{\sqrt{g}} V^{\mu} \partial_{\mu} \sqrt{g} \equiv \frac{1}{\sqrt{g}} \partial_{\mu} \left( V^{\mu} \sqrt{g} \right). \tag{18.7}$$

That is, the expression for the covariant divergence is charmingly simple.

# The divergence theorem

Under coordinate transformations, the volume element changes like

$$d^{n}x \to d^{n}\overline{x} \det\left(\frac{\partial x}{\partial \overline{x}}\right)$$
(18.8)

where det  $\left(\frac{\partial x}{\partial \overline{x}}\right)$  is the Jacobian of the transformation.

But

$$\overline{g}_{\mu\nu} = g_{\kappa\lambda} \frac{\partial x^{\kappa}}{\partial \overline{x}^{\mu}} \frac{\partial x^{\lambda}}{\partial \overline{x}^{\nu}}$$
(18.9)

so, using a well-known property of determinants of matrix-products,

$$\det(AB) = \det(A) \det(B), \qquad (18.10)$$

we find

$$\overline{g} = g \left[ \det \left( \frac{\partial x}{\partial \overline{x}} \right) \right]^2$$
(18.11)

i.e.

$$\sqrt{g} d^n x = \sqrt{\overline{g}} d^n \overline{x} \,. \tag{18.12}$$

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In other words,  $\sqrt{g} d^n x$  is the invariant volume element in a general space of *n* dimensions.

Now we can transform the volume integral of a divergence into an integral of a vector over a (normal) hypersurface.

$$\int \operatorname{div} V \sqrt{g} \, d^n x = \int V^{\mu}_{;\,\mu} \sqrt{g} \, d^n x = \int d^n x \, \partial_{\mu} \left( \sqrt{g} \, V^{\mu} \right) \equiv \int dS_{\mu} \, V^{\mu} \, \sqrt{g} \,. \tag{18.13}$$

## Proof of the theorem about determinants:

We want to prove that for some matrix G,

 $\log \left[ \det(G) \right] = \operatorname{Tr} \left[ \log(G) \right].$ 

Now this is obvious if the matrix is diagonalizable, with eigenvalues  $g_k$  since then

$$\log \left[\det(G)\right] = \log \left[\prod_{k=1}^{N} g_{k}\right] \equiv \sum_{k=1}^{N} \log(g_{k})$$

and

$$\operatorname{Tr} \left[ \log(G) \right] = \sum_{k=1}^{N} \log(g_k) \,.$$

We are concerned to prove the theorem more generally. First, it had better be true that the matrix

$$A = \log(G)$$

exists (that is, it can be defined, the matrix has no zero eigenvalues, *etc. etc.*). Assuming this is the case, let

$$G(\lambda) \stackrel{df}{=} e^{\lambda A}, \qquad G(1) = e^{A} = G.$$

Now let us define

$$d \log \left[\det(G(\lambda))\right] \stackrel{df}{=} \log \left[\det(G(\lambda + d\lambda))\right] - \log \left[\det(G(\lambda))\right]$$
$$= \log \left[\det(G + dG)\right] - \log \left[\det(G)\right]$$
$$= \log \left[\det(G(1 + G^{-1}dG))\right] - \log \left[\det(G)\right]$$
$$= \log \left[\det(1 + G^{-1}dG)\right] .$$

Now let us compute the last term:

Proof of the theorem about determinants:

$$\det(1+G^{-1}dG) = \begin{vmatrix} 1+(G^{-1}dG)_{11} & (G^{-1}dG)_{12} & (G^{-1}dG)_{13} & \dots \\ (G^{-1}dG)_{21} & 1+(G^{-1}dG)_{22} & (G^{-1}dG)_{23} & \dots \\ (G^{-1}dG)_{31} & (G^{-1}dG)_{32} & 1+(G^{-1}dG)_{33} & \dots \\ \dots & \dots & \dots & \dots \end{vmatrix}$$
$$= \prod_{k} \left[ 1+(G^{-1}dG)_{kk} \right] + O\left((G^{-1}dG)^{2}\right)$$
$$\approx 1 + \sum_{k} \left(G^{-1}dG\right)_{kk} \equiv 1 + \operatorname{Tr}\left(G^{-1}dG\right).$$

To this same order, then,

$$d\log\left[\det(G(\lambda))\right] = \log\left[1 + \operatorname{Tr}\left(G^{-1}dG\right)\right] = \operatorname{Tr}\left(G^{-1}dG\right)$$

However, since  $G(\lambda) = e^{\lambda A}$ , clearly  $dG(\lambda) = Ae^{\lambda A} d\lambda$ 

and thus

 $d \log [\det(G(\lambda))] = \operatorname{Tr}(G^{-1}dG) = \operatorname{Tr}(e^{-\lambda A} A e^{\lambda A}) d\lambda = \operatorname{Tr}(A) d\lambda$ , giving, by direct integration,

 $\log \left[ \det(G(\lambda)) \right] \; = \; \lambda \operatorname{Tr}(A) \; + \; \operatorname{constant} \; .$ 

Since both sides must vanish when  $\lambda = 0$ , the constant is zero, giving at last (with  $\lambda = 1$ ) det(G) = exp [Tr(A)] = exp [Tr(log(G))].