> with(DifferentialGeometry):
 with(Tensor):

First we initialize the base manifold.

Here is the infinitesimal generator of the symmetry: a vector field on M.

An image is modeled as a smooth section of a Fibered Manifold. Initializing the FM:

We will need to differentiate smooth sections, so here is a (zero) connection:

Color > Gamma3:=eval(Connection([nabla(D_x,D_R)=evalDG(a*D_R)]), a = 0);
$$\Gamma 3 := \nabla_{\partial_x} \partial_R = 0 \partial_R$$
 (4)

An image is a smooth section of the fiber bundle.

Color > image:=evalDG(f1(x,y)*D_R + f2(x,y)*D_G + f3(x,y)* D_B);

$$image := f1(x, y) \partial_R + f2(x, y) \partial_G + f3(x, y) \partial_B$$
 (5)

Clumsy code: we need Maple to recognize that the vector field lives in the base space.

Color > symC:=subs(M=Color,sym);
$$symC := g1(x, y) \partial_x + g2(x, y) \partial_y$$
 (6)

The classification model can be seen as a function of the image with 10 outputs (10 classes).

As before, we define a (zero) connection for later differentiation.

Here are the coefficients of the model: 10 functions of an image.

EN > funcs:=[seq(1/(1+exp(-F||i(R,G,B))),i=1..10)];
funcs:=
$$\left[\frac{1}{1+e^{-FI(R,G,B)}}, \frac{1}{1+e^{-F2(R,G,B)}}, \frac{1}{1+e^{-F3(R,G,B)}}, \frac{1}{1+e^{-F4(R,G,B)}}, \frac{1}{1+e^{-F4(R,G,B)}}, \frac{1}{1+e^{-F5(R,G,B)}}, \frac{1}{1+e^{-F6(R,G,B)}}, \frac{1}{1+e^{-F7(R,G,B)}}, \frac{1}{1+e^{-F8(R,G,B)}}, \frac{1}{1+e^{-F8(R,G,B)}}, \frac{1}{1+e^{-F10(R,G,B)}}\right]$$

Below is the derivative of the model with respect to the image.

Preparing to realize this as a function of x and y:

EN > eqns:=[R=f1(x,y),G=f2(x,y),B=f3(x,y)];

$$eqns := [R = f1(x,y), G = f2(x,y), B = f3(x,y)]$$
 (10)

Now realizing it as a function of x and y.

EN > Grad:=DEtools:-dsubs(eqns,grad):

The chain rule applies: we have to multiply by the derivative of the image w.r.t. x and y.

EN > dd:=DirectionalCovariantDerivative(symC,image,Gamma3);
$$dd := \left(g1(x,y) \left(\frac{\partial}{\partial x} f1(x,y)\right) + g2(x,y) \left(\frac{\partial}{\partial y} f1(x,y)\right)\right) \partial_R + \left(g1(x,y) \left(\frac{\partial}{\partial x} f2(x,y)\right) + g2(x,y) \left(\frac{\partial}{\partial y} f2(x,y)\right)\right) \partial_G + \left(g1(x,y) \left(\frac{\partial}{\partial x} f3(x,y)\right) + g2(x,y) \left(\frac{\partial}{\partial y} f3(x,y)\right)\right) \partial_R + \left(g1(x,y) \left(\frac{\partial}{\partial y} f3(x,y)\right) + g2(x,y) \left(\frac{\partial}{\partial y} f3(x,y)\right)\right) \partial_R + \left(g1(x,y) \left(\frac{\partial}{\partial y} f3(x,y)\right) + g2(x,y) \left(\frac{\partial}{\partial y} f3(x,y)\right)\right) \partial_R + \left(g1(x,y) \left(\frac{\partial}{\partial y} f3(x,y)\right) + g2(x,y) \left(\frac{\partial}{\partial y} f3(x,y)\right)\right) \partial_R + \left(g1(x,y) \left(\frac{\partial}{\partial y} f3(x,y)\right) + g2(x,y) \left(\frac{\partial}{\partial y} f3(x,y)\right)\right) \partial_R + \left(g1(x,y) \left(\frac{\partial}{\partial y} f3(x,y)\right) + g2(x,y) \left(\frac{\partial}{\partial y} f3(x,y)\right)\right) \partial_R + \left(g1(x,y) \left(\frac{\partial}{\partial y} f3(x,y)\right) + g2(x,y) \left(\frac{\partial}{\partial y} f3(x,y)\right)\right) \partial_R + g2(x,y) \partial_R +$$

Another clumsy coding moment: converting this to the base space of the EN vector

bundle.

Color > DD:=DGzip(DGinformation(dd, "CoefficientList", "all"), DGinformation(EN, "FrameBaseVectors"), "plus");
$$DD := \left(g1(x,y) \left(\frac{\partial}{\partial x} f1(x,y)\right) + g2(x,y) \left(\frac{\partial}{\partial y} f1(x,y)\right)\right) \partial_R + \left(g1(x,y) \left(\frac{\partial}{\partial x} f2(x,y)\right) + g2(x,y) \left(\frac{\partial}{\partial y} f2(x,y)\right)\right) \partial_G + \left(g1(x,y) \left(\frac{\partial}{\partial x} f3(x,y)\right) + g2(x,y) \left(\frac{\partial}{\partial y} f3(x,y)\right)\right) \partial_B$$

$$f3(x,y) \right) \partial_B$$

Now we "multiply" the terms together.

_EN > ans2:=ContractIndices(Grad,DD,[[2,1]]):

Is this the right answer? let's check. The model is a smooth section of pi:R10 -> R2.

Once again, we will need to differentiate, so we define a (zero) connection.

N > Gamma:=eval(Connection([nabla(D_x,E1)=a*E1]),a=0);
$$\Gamma := \nabla_{\underset{x}{0}} E1 = 0 E1$$
 (15)

The functions given before are, at the core, functions of x and y:

Here is the neural network. The functions F are learned during training.

$$\begin{array}{l} \textbf{N > Model:=DGzip(Funcs,DGinformation(N,"FrameFiberVectors"),} \\ \textbf{"plus");} \\ Model := \frac{1}{1+e^{-FI(fI(x,y),\,f2(x,y),\,f3(x,y))}} \,\,E1 + \frac{1}{1+e^{-F2(fI(x,y),\,f2(x,y),\,f3(x,y))}} \,\,E2 \\ + \frac{1}{1+e^{-F3(fI(x,y),\,f2(x,y),\,f3(x,y))}} \,\,E3 + \frac{1}{1+e^{-F4(fI(x,y),\,f2(x,y),\,f3(x,y))}} \,\,E4 \\ + \frac{1}{1+e^{-F5(fI(x,y),\,f2(x,y),\,f3(x,y))}} \,\,E5 + \frac{1}{1+e^{-F6(fI(x,y),\,f2(x,y),\,f3(x,y))}} \,\,E6 \\ + \frac{1}{1+e^{-F7(fI(x,y),\,f2(x,y),\,f3(x,y))}} \,\,E7 + \frac{1}{1+e^{-F8(fI(x,y),\,f2(x,y),\,f3(x,y))}} \,\,E8 \\ + \frac{1}{1+e^{-F9(fI(x,y),\,f2(x,y),\,f3(x,y))}} \,\,E9 + \frac{1}{1+e^{-F10(fI(x,y),\,f2(x,y),\,f3(x,y))}} \,\,E10 \end{array}$$

Awkward coding again: need to recognize the vector field as living on the base space of N.

N > symCn:=subs(Color=N,symC);

$$symCn := g1(x, y) \partial_x + g2(x, y) \partial_y$$
(18)

This is the real answer:

N > ans:=DirectionalCovariantDerivative(symCn,Model,Gamma):

Is it equal to what we got using the product rule? We can check, but there's one last awkward coding moment: realizing our previous answer as a section of the bundle N (which is equivalent to EN).

The quantities are equal, meaning the directional covariant derivative of the model along a vector field can be computed using the product rule.