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# Derivatives of the G matrix

## Helpful Notes

When reading *Wilson, Decius, & Cross* (WDC) the authors introduce some odd terminology at least according to modern notation. First off, they define their internal coordinates by  $S_i$  which, like, fine. I'm gonna use  $r_i$

Then given a reference structure which I'll call  $\mathbf{S}_e$  and whose existence they allow to be implicit, they approximate the internals up to first order by Cartesians by an equation that they write as

$$S_t = \sum_{i=1}^N B_{ti} \xi_i, \quad t = 1, 2, \dots, M$$

where  $N$  and  $M$  are the number of Cartesian coordinates and number of respective internal coordinates

Here we have  $\xi_i$  being the Cartesian displacement. I'd usually call it  $\Delta x_i$ . And the  $B$  term is just the derivative of the internal with respect to the Cartesians. This can be much stated more cleanly by saying that we'll represent  $S_t$  as

$$S_t: \left( \frac{\partial S_t}{\partial x_1} \Delta x_1, \frac{\partial S_t}{\partial x_2} \Delta x_2, \dots, \frac{\partial S_t}{\partial x_{3N}} \Delta x_{3N} \right) = \left( \frac{\partial S_t}{\partial x_i} \Delta x_i \right)_{i=1, \dots}$$

This is saying that we'll linearly approximate our internal as a Cartesian displacement. We won't actually use this definition to evaluate the our internals, but these derivatives are important as this is how we get the  $B$  matrix that is implicitly referred to in the text

$$B = \begin{pmatrix} \frac{\partial S_1}{\partial x_1} & \frac{\partial S_1}{\partial x_2} & \cdots & \frac{\partial S_1}{\partial x_N} \\ \frac{\partial S_2}{\partial x_1} & \frac{\partial S_2}{\partial x_2} & & \\ \vdots & & \ddots & \vdots \\ \frac{\partial S_M}{\partial x_1} & & \cdots & \frac{\partial S_M}{\partial x_N} \end{pmatrix}$$

For anyone who's taken intro calc or ever transformed from Cartesian to polar coordinates to do an integral, this is just the Jacobian matrix and I'm not sure why *WDC* don't refer to it as such. Returning to the notation that I prefer (and introducing a slightly odd gradient notation), we have

$$\nabla_X R = \begin{pmatrix} \frac{\partial r_1}{\partial x_1} & \frac{\partial r_1}{\partial x_2} & \cdots & \frac{\partial r_1}{\partial x_N} \\ \frac{\partial r_2}{\partial x_1} & \frac{\partial r_2}{\partial x_2} & & \\ \vdots & & \ddots & \vdots \\ \frac{\partial r_M}{\partial x_1} & & \cdots & \frac{\partial r_M}{\partial x_N} \end{pmatrix}$$

Next, in this notation we're given that the G-matrix is defined as

$$G_{ij} = \sum_{\alpha=1}^N \frac{1}{m_\alpha} \frac{\partial}{\partial x_\alpha} r_i \frac{\partial}{\partial x_\alpha} r_j$$

This is all well and good, but we can write this in a more intuitive way. Letting

$$\nabla_X r_i = \left( \frac{\partial r_i}{\partial x_\alpha} \Delta x_\alpha \right)_{\alpha=1, \dots, 3N}$$

we'll define a mass-weighted form of this by

$$\nabla_Y r_i = \left( \frac{1}{\sqrt{m_\alpha}} \frac{\partial r_i}{\partial x_\alpha} \Delta x_\alpha \right)_{\alpha=1, \dots, 3N}$$

the notation for which is motivated by the mass-weighted Coordinate transformation

$$y_\alpha = \sqrt{m_\alpha} x_\alpha$$

taking all of this together we get

$$G_{ij} = \nabla_Y r_i \cdot \nabla_Y r_j$$

This type of definition of a matrix element (vector<sub>i</sub> · vector<sub>j</sub>) tells us that our matrix can be constructed by a matrix multiplication. In this case we have

$$G = (\nabla_Y R) (\nabla_Y R)^T$$

where we can build  $\nabla_Y R$  by

$$\nabla_Y R = (\nabla_X R) M^{-\frac{1}{2}}$$

with

$$M = \begin{pmatrix} m_1 & & & \\ & m_2 & & \\ & & \ddots & \\ & & & m_{3N} \end{pmatrix}$$

## Derivation

Since we use it so often, we let the mass-weighted Jacobian be

$$J = \nabla_Y R$$

With this, the G matrix is given

$$G=JJ^T$$

Next let's say we want the derivatives of the G matrix with respect to a new set of coordinates,  $\{Q_n\}$ .

Going back to prior notes on tensor derivatives, we had

$$\nabla_Q V = \nabla_Q x \odot \nabla_x V$$

where V was our potential function in Cartesian coordinates, Q was our internal coordinate set, and  $\odot$  represents a generalized matrix multiplication to tensors (implemented in **numpy** via **tensordot**).

Correspondingly since J is defined in our R coordinates, this means that

$$\nabla_Q J = \nabla_Q R \odot \nabla_R J$$

the  $\nabla_Q R$  term is the Jacobian for going from the Q coordinate set to the R coordinate set. In the special case the  $Q_i$  are defined as linear combinations of the R coordinates defined by some transformation matrix L, this reduces to being simply

$$\nabla_Q R = L^{-1}$$

Next we have the  $\nabla_R J$  term, a three dimensional tensor whose elements look like

$$(\nabla_R J)_{ijk} = \frac{\partial}{\partial r_i} \left( \frac{\partial r_j}{\partial x_k} \right) = \frac{\partial r_j^2}{\partial r_i \partial x_k} = \frac{\partial}{\partial x_k} \left( \frac{\partial r_j}{\partial r_i} \right) = \frac{\partial r_j}{\partial x_k} \delta_{ij}$$

which means this is just a nice block-diagonal tensor. This fact can be convenient when implementing this.

## G-matrix Derivatives

### First Derivative Tensor

Next, moving to the full term we have

$$\begin{aligned} \nabla_Q G &= \nabla_Q (JJ^T) \\ &= (\nabla_Q J) \odot J^T + (J \odot (\nabla_Q J^T)^{2 \rightarrow 1})^{2 \rightarrow 1} \end{aligned}$$

Next, we'll note that this can even be reduced somewhat further by considering that

$$\begin{aligned} \nabla_Q J^T &= \nabla_Q (J^{2 \rightarrow 1}) \\ &= (\nabla_Q J)^{3 \rightarrow 2} \end{aligned}$$

and then

$$\begin{aligned} ((\nabla_Q J^T)^{2 \rightarrow 1}) &= ((\nabla_Q J)^{3 \rightarrow 2})^{2 \rightarrow 1} \\ &= (\nabla_Q J)^{3 \rightarrow 1} \end{aligned}$$

we have

$$\nabla_Q G = (\nabla_Q J) \odot J^T + (J \odot (\nabla_Q J)^{3 \rightarrow 1})^{2 \rightarrow 1}$$

which is nice as this allows the tensor  $\nabla_Q J$  to be reused

For ease of reading, we'll introduce a compact notation

$$\begin{aligned} G_Q &= \nabla_Q G \\ &= J_Q J^{2:1} + (J J_Q^{3:1})^{2:1} \end{aligned}$$

## Validation

### Second Derivative Tensor

We also need second derivatives of our G-matrix. To get this we'll evaluate

$$\begin{aligned} \nabla_Q^2 G &= \nabla_Q (\nabla_Q G) \\ &= \nabla_Q ((\nabla_Q J) \odot J^T + (J \odot (\nabla_Q J)^{3 \rightarrow 1})^{2 \rightarrow 1}) \end{aligned}$$

or in compact notation:

$$\begin{aligned} G_{Q^2} &= (J_Q J^{2:1} + (J J_Q^{3:1})^{2:1})_Q \\ &= (J_Q J^{2:1})_Q + ((J J_Q^{3:1})^{2:1})_Q \end{aligned}$$

then taking the first term we'll let

$$\begin{aligned} A &= J_Q \\ B &= J^{2:1} \end{aligned}$$

$$\begin{aligned} (J_Q J^{2:1})_Q &= (AB)_Q \\ &= A_Q B + (A (B_Q)^{2:1})^{a:1} \\ &= (J_Q)_Q J^{2:1} + (J_Q ((J^{2:1})_Q)^{2:1})^{3:1} \\ &= (J_{Q^2} J^{2:1}) + (J_Q (J_Q^{3:2})^{2:1})^{3:1} \\ &= (J_{Q^2} J^{2:1}) + (J_Q J_Q^{3:1})^{3:1} \end{aligned}$$

and then for the second we'll let

$$\begin{aligned} A &= J \\ B &= J_Q^{3:1} \end{aligned}$$

$$\begin{aligned} ((AB)^{2:1})_Q &= ((AB)_Q)^{3:2} \\ &= (A_Q B + (A (B_Q)^{2:1})^{a:1})^{3:2} \\ &= (A_Q B)^{3:2} + (A (B_Q)^{2:1})^{1:3} \\ &= (J_Q J_Q^{3:1})^{3:2} + (J (J_Q^{4:2})^{2:1})^{1:3} \\ &= (J_Q J_Q^{3:1})^{3:2} + (J J_Q^{4:1})^{1:3} \end{aligned}$$

stitching these together we have

$$\nabla_{Q^2} G = (J_{Q^2} J^{2:1}) + (J_Q J_Q^{3:1})^{3:1} + (J_Q J_Q^{3:1})^{3:2} + (J J_{Q^2}^{4:1})^{1:3}$$

### Validation

Finally we'll check our terms

$$\begin{aligned} (J_{Q^2} J^{2:1})_{ijkl} &= (J_{Q^2})_{ijk:} (J^{2:1})_{:l} \\ &= (J_{Q^2})_{ijk:} (J_l) \\ &= \left( \left( \frac{\partial^2}{\partial Q_i \partial Q_j} \frac{\partial r_k}{\partial y_\alpha} \right)_{\alpha=1, \dots, N} \right) \cdot \left( \left( \frac{\partial r_l}{\partial y_\alpha} \right)_{\alpha=1, \dots, N} \right) \\ &= \sum_{\alpha=1}^N \frac{1}{m_\alpha} \frac{\partial^3 r_k}{\partial Q_i \partial Q_j \partial x_\alpha} \frac{\partial r_l}{\partial x_\alpha} \\ ((J_Q (J_Q)^{3:1})^{3:1})_{ijkl} &= (J_Q (J_Q)^{3:1})_{jkil} \\ &= (J_Q)_{jk:} ((J_Q)^{3:1})_{:il} \\ &= (J_Q)_{jk:} (J_Q)_{il:} \\ &= \left( \left( \frac{\partial}{\partial Q_j} \frac{\partial r_k}{\partial y_\alpha} \right)_{\alpha=1, \dots, N} \right) \cdot \left( \left( \frac{\partial}{\partial Q_i} \frac{\partial r_l}{\partial y_\alpha} \right)_{\alpha=1, \dots, N} \right) \\ &= \sum_{\alpha=1}^N \frac{1}{m_\alpha} \frac{\partial}{\partial Q_j} \frac{\partial r_k}{\partial x_\alpha} \frac{\partial}{\partial Q_i} \frac{\partial r_l}{\partial x_\alpha} \\ ((J_Q (J_Q)^{3:1})^{3:2})_{ijkl} &= (J_Q (J_Q)^{3:1})_{ikjl} \\ &= (J_Q)_{ik:} ((J_Q)^{3:1})_{:jl} \\ &= (J_Q)_{ik:} (J_Q)_{jl:} \\ &= \left( \left( \frac{\partial}{\partial Q_i} \frac{\partial r_k}{\partial y_\alpha} \right)_{\alpha=1, \dots, N} \right) \cdot \left( \left( \frac{\partial}{\partial Q_j} \frac{\partial r_l}{\partial y_\alpha} \right)_{\alpha=1, \dots, N} \right) \\ &= \sum_{\alpha=1}^N \frac{1}{m_\alpha} \frac{\partial}{\partial Q_i} \frac{\partial r_k}{\partial x_\alpha} \frac{\partial}{\partial Q_j} \frac{\partial r_l}{\partial x_\alpha} \\ ((J J_{Q^2}^{4:1})^{1:3})_{ijkl} &= (J J_{Q^2}^{4:1})_{kijl} \\ &= J_{k:} (J_{Q^2}^{4:1})_{:ijl} \\ &= J_{k:} (J_{Q^2})_{ijl:} \\ &= \left( \left( \frac{\partial r_k}{\partial y_\alpha} \right)_{\alpha=1, \dots, N} \right) \cdot \left( \left( \frac{\partial^2}{\partial Q_i \partial Q_j} \frac{\partial r_l}{\partial y_\alpha} \right)_{\alpha=1, \dots, N} \right) \\ &= \sum_{\alpha=1}^N \frac{1}{m_\alpha} \frac{\partial r_k}{\partial x_\alpha} \frac{\partial^3 r_l}{\partial Q_i \partial Q_j \partial x_\alpha} \end{aligned}$$

Now keeping in mind that we want

$$(G_{Q^2})_{ijkl} = \frac{\partial^2}{\partial Q_i \partial Q_j} \left( \sum_{\alpha=1}^N \frac{1}{m_\alpha} \frac{\partial r_k}{\partial x_\alpha} \frac{\partial r_l}{\partial x_\alpha} \right)$$

$$\begin{aligned}
&= \frac{\partial}{\partial Q_i} \sum_{\alpha=1}^N \frac{1}{m_\alpha} \left( \frac{\partial^2 r_k}{\partial Q_j \partial x_\alpha} \frac{\partial r_l}{\partial x_\alpha} + \frac{\partial r_k}{\partial x_\alpha} \frac{\partial^2 r_l}{\partial Q_j \partial x_\alpha} \right) \\
&= \sum_{\alpha=1}^N \frac{1}{m_\alpha} \left( \frac{\partial^3 r_k}{\partial Q_i \partial Q_j \partial x_\alpha} \frac{\partial r_l}{\partial x_\alpha} + \frac{\partial^2 r_k}{\partial Q_j \partial x_\alpha} \frac{\partial^2 r_l}{\partial Q_i \partial x_\alpha} + \frac{\partial^2 r_k}{\partial Q_i \partial x_\alpha} \frac{\partial^2 r_l}{\partial Q_j \partial x_\alpha} + \frac{\partial r_k}{\partial x_\alpha} \frac{\partial^3 r_l}{\partial Q_i \partial Q_j \partial x_\alpha} \right)
\end{aligned}$$

we see this has worked as desired

### Relationship to Cartesians

We saw previously that we had

$$\mathbf{J}_Q = \nabla_Q \mathbf{J} = \nabla_Q \mathbf{Y} \odot \nabla_Y \mathbf{J} = \mathbf{Y}_Q \mathbf{J}_Y$$

Now we'll consider the second derivative form of this

$$\begin{aligned}
\nabla_{Q^2} \mathbf{J} &= \mathbf{J}_{Q^2} \\
&= (\mathbf{Y}_Q \mathbf{J}_Y)_Q \\
&= \mathbf{Y}_{Q^2} \mathbf{J}_Y + (\mathbf{Y}_Q ((\mathbf{J}_Y)_Q)^{2:1})^{2:1} \\
&= \mathbf{Y}_{Q^2} \mathbf{J}_Y + (\mathbf{Y}_Q (\mathbf{Y}_Q \mathbf{J}_{Y^2})^{2:1})^{2:1}
\end{aligned}$$

What if our Q coordinates are already linear combinations of Cartesians? In that case, calling this transformation L, we have

$$\begin{aligned}
\mathbf{J} &= \mathbf{L} \mathbf{M}^{-\frac{1}{2}} \\
\mathbf{J}_Y &= \mathbf{O} \\
\mathbf{J}_{Y^2} &= \mathbf{O}
\end{aligned}$$

Alternately, what if we our Q coordinates to be linear combinations of internals (R), but we have some set of modes  $\mathbf{Q}^{\text{Cart}}$  that are already defined as linear combinations of Cartesians?

In this case we have, calling that transformation L,

$$\mathbf{J} = \mathbf{L} \mathbf{M}^{-\frac{1}{2}} \nabla_X \mathbf{R}$$

(note that if  $\mathbf{R} = \mathbf{X}$ , this reduces to the previous case)

Then we have

$$\mathbf{J}_Y = (\mathbf{L} \mathbf{M}^{-\frac{1}{2}} \mathbf{R}_X)_Y = \mathbf{L} \mathbf{M}^{-\frac{1}{2}} (\mathbf{R}_X)_Y = \mathbf{L} \mathbf{M}^{-\frac{1}{2}} (\mathbf{X}_Y \mathbf{R}_{X^2})$$

keeping in mind what  $\mathbf{X}_Y$  really is

$$\mathbf{X}_Y = \begin{pmatrix} \frac{\partial x_1}{\partial y_1} & \frac{\partial x_1}{\partial y_2} & \cdots & \frac{\partial x_1}{\partial y_N} \\ \frac{\partial x_2}{\partial y_1} & \frac{\partial x_2}{\partial y_2} & & \\ \vdots & & \ddots & \vdots \\ \frac{\partial x_N}{\partial y_1} & & \cdots & \frac{\partial x_N}{\partial y_N} \end{pmatrix} = \begin{pmatrix} \frac{1}{\sqrt{m_1}} & & & \\ & \frac{1}{\sqrt{m_2}} & & \\ & & \ddots & \\ & & & \frac{1}{\sqrt{m_N}} \end{pmatrix} = \mathbf{M}^{-\frac{1}{2}}$$

we get

$$\mathbf{J}_Y = \mathbf{L} \mathbf{M}^{-1} \mathbf{R}_{X^2}$$

and

$$\mathbf{J}_{Y^2} = \left( \mathbf{L} \mathbf{M}^{-\frac{1}{2}} \mathbf{R}_X \right)_Y = \left( \mathbf{L} \mathbf{M}^{-1} \mathbf{R}_{X^2} \right)_Y = \mathbf{L} \mathbf{M}^{-\frac{3}{2}} \mathbf{R}_{X^3}$$

Then we have the  $\mathbf{Y}_Q$  terms, which will be slightly funky

$$\mathbf{Y}_Q = \mathbf{R}_Q \mathbf{Y}_R = \mathbf{J}^{-1} (\mathbf{X}_R \mathbf{Y}_X) = \mathbf{J}^{-1} \mathbf{X}_R \mathbf{M}$$

these  $\mathbf{X}_R$  can either be computed or by finding the pseudoinverse of  $\mathbf{R}_X$

Finally

$$\begin{aligned} \mathbf{Y}_{Q^2} &= (\mathbf{R}_Q \mathbf{X}_R \mathbf{M})_Q \\ &= \mathbf{R}_{Q^2} \mathbf{X}_R \mathbf{M} + (\mathbf{R}_Q (\mathbf{X}_R)_Q^{2:1})^{2:1} \mathbf{M} \\ &= (\mathbf{R}_{Q^2} \mathbf{X}_R + (\mathbf{R}_Q (\mathbf{R}_Q \mathbf{X}_{R^2})^{2:1})^{2:1}) \mathbf{M} \end{aligned}$$

then we note that  $\mathbf{R}_{Q^2} = \mathbf{O}$  since we have a linear transform, giving us

$$\mathbf{Y}_{Q^2} = (\mathbf{J}^{-1} (\mathbf{J}^{-1} \mathbf{X}_{R^2})^{2:1})^{2:1} \mathbf{M}$$