

# The Hessian of a Cost Function

Herman Jaramillo

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## 1 The problem

Given the least square cost function

$$\mathcal{C}(p) = \frac{1}{2} \langle u(p) - d, u(p) - d \rangle$$

This is pure algorithmic. we do not define the Hessian rigorously.

Pratt et. al., [1] show a compact representation of FWI problem using matrices and in their representation they find ways to compute the Hessian.

Here we formulate the problem in terms of functionals and provide new insights into details about the computations.

The Hessian is the derivative of the gradient. The gradient is

$$\nabla_p(\mathcal{C}(p)) = \langle u_p, u - d \rangle$$

assuming that we are working over the real spaces.

Then

$$(\nabla_p \mathcal{C}(p))_p = \langle u_p, u_p \rangle + \langle (u_p)_p, u - d \rangle.$$

Pratt et. al., call  $u_d = J$  where  $J$  is for Jacobian. In matrix notation this is

$$H = (\nabla_p \mathcal{C}(p))_p = J^T J + (dJ/dp)^T (u - d) = H_1 + H_2.$$

The first term ( $H_1$ ) is used as an approximation of the Gauss–Newton iteration technique, while both terms ( $H$ ) provide the full Newton inversion. If  $u = Ap$ , then  $J = A$  and  $\nabla_p(\mathcal{C}(p))_p = A^T A$ . This is the reason why

sometimes abuse the language saying that  $A^T A$  is the Hessian, for the least square approximation of  $u \approx Ap$ .

Let us assume we know  $u_p$ . Then  $H_1$  is easy to compute as

$$H_1 = J^T J.$$

Now, for  $H_2$ , starting at the modeling equation:

$$A(p)u = f.$$

Take first derivative

$$A_p u + A u_p = 0,$$

and second derivative

$$(A_p)_p u + A_p u_p + A_p u_p + A(u_p)_p = 0,$$

so

$$A(u_p)_p = -(A_p)_p u - 2A_p u_p$$

As in the gradient computation, let us choose  $v$  such that

$$A^* v = u - d,$$

(the same already obtained by the gradient method) so

$$\langle A(u_p)_p, v \rangle = \langle (u_p)_p, u - d \rangle = H_2,$$

and the second part of the Hessian is computed with the equation

$$\begin{aligned} H_2 &= \langle -(A_p)_p u - 2A_p u_p, v \rangle \\ &= -\langle (A_p)_p u, v \rangle - 2\langle A_p u_p, v \rangle \\ &= -\langle u, [(A_p)_p]^* v \rangle - 2\langle u_p, [A_p]^* v \rangle \end{aligned}$$

But, again, this needs that we know  $u_p$  and the computation of  $u_p$  is very expensive.

## References

- [1] Pratt R. G., Shin G., and Hicks G. J. Gauss–newton and full newton methods in frequency–space seismic waveform inversion. *Geophys. J. Int.*, (133):341–362, 1998.