

**Newton-Raphson step.** Quadratic expansion

$$E \approx E_0 + \Delta \mathbf{x}^t \mathbf{g}_0 + \frac{1}{2} \Delta \mathbf{x}^t \mathbf{H}_0 \Delta \mathbf{x} \quad \Delta \mathbf{x} \equiv \mathbf{x} - \mathbf{x}_0$$

Linear gradient expansion

$$\mathbf{g} = \mathbf{g}_0 + \mathbf{H} \Delta \mathbf{x}_0$$

Setting the gradient at the next point to zero gives the Newton-Raphson step

$$\mathbf{g} \stackrel{!}{=} \mathbf{0} \implies \Delta \mathbf{x} = -\mathbf{H}_0^{-1} \mathbf{g}_0$$

For a perfectly quadratic surface, the Hessian is constant and the Newton-Raphson step takes us directly to the stationary point. On a surface with cubic and higher-order terms, this step can be repeated iteratively until we are close enough to the stationary point that the region separating us from it is approximately quadratic.

**The secant condition.** For points  $\mathbf{x}$  and  $\mathbf{x}_0$  sharing a locally quadratic region with each other, the change in the gradient between them is described by the following.

$$\mathbf{H} \Delta \mathbf{x} \approx \mathbf{H}_0 \Delta \mathbf{x} \approx \Delta \mathbf{g} \quad \Delta \mathbf{g} \equiv \mathbf{g} - \mathbf{g}_0$$

This can be used to determine an approximation  $\tilde{\mathbf{H}}$  to the Hessian at  $\mathbf{x}$  using the Hessian at  $\mathbf{x}_0$ . Namely, we require the approximation to satisfy

$$\tilde{\mathbf{H}} \Delta \mathbf{x} \stackrel{!}{=} \Delta \mathbf{g} \quad \tilde{\mathbf{H}} = \mathbf{H}_0 + \Delta \tilde{\mathbf{H}}$$

which is known as the *quasi-Newton condition*, or alternatively the *secant condition*. If the dimension is  $d$ , then we have  $d$  linear equations and  $d^2$  undetermined matrix elements (or rather  $d(d+1)/2$  matrix elements, by symmetry), so the system is underdetermined – we have an infinite number of solutions.

The simplest way forward is to assume  $\Delta \tilde{\mathbf{H}}$  has rank 1, which for a symmetric matrix implies that it has the form

$$\Delta \tilde{\mathbf{H}} = \eta \mathbf{e} \mathbf{e}^t$$

where  $\eta$  is a scalar and  $\mathbf{e}$  is a unit vector. This is the *symmetric rank-1* (SR1) approximation. Substituting this into the secant condition, we find

$$\mathbf{e} = \frac{\Delta \mathbf{g} - \mathbf{H}_0 \Delta \mathbf{x}}{\|\Delta \mathbf{g} - \mathbf{H}_0 \Delta \mathbf{x}\|} \quad \eta = \frac{\mathbf{e} \cdot (\Delta \mathbf{g} - \mathbf{H}_0 \Delta \mathbf{x})}{\mathbf{e} \cdot \Delta \mathbf{x}} = \frac{(\Delta \mathbf{g} - \mathbf{H}_0 \Delta \mathbf{x})^2}{(\Delta \mathbf{g} - \mathbf{H}_0 \Delta \mathbf{x}) \cdot \Delta \mathbf{x}}$$

which allows us to simplify the SR1 matrix update.

$$\Delta \tilde{\mathbf{H}}_{\text{SR1}} \equiv \frac{(\Delta \mathbf{g} - \mathbf{H}_0 \Delta \mathbf{x})(\Delta \mathbf{g} - \mathbf{H}_0 \Delta \mathbf{x})^t}{(\Delta \mathbf{g} - \mathbf{H}_0 \Delta \mathbf{x}) \cdot \Delta \mathbf{x}}$$

This provides a scheme for updating the Hessian using the change in the first derivative as we move across the potential surface. If we take several steps in a small, locally quadratic region of the surface, this approximation will improve with each step, and eventually we will have quite a good approximation to  $\mathbf{H}$ . We can determine other approximations by assuming that the Hessian correction has rank 2, which can be used to derive the *Broyden-Fletcher-Goldfarb-Shanno* (BFGS) and *Davidon-Fletcher-Powell* (DFP) Hessian updates.

Combining secant updating with Newton-Raphson optimization leads to the *quasi-Newton optimization algorithms*, which have the following form.

1. Evaluate the gradient  $\mathbf{g}_0$  at the starting point  $\mathbf{x}_0$  and either evaluate the Hessian or use an approximation. Setting  $\tilde{\mathbf{H}}_0 = \frac{\|\mathbf{g}_0\|}{s_0} \mathbf{1}$  will yield a gradient descent of length  $s_0$  for the initial step.

2. Take a Newton-Raphson step.

$$\mathbf{x} = \mathbf{x}_0 + \Delta\mathbf{x} \quad \Delta\mathbf{x} = -\mathbf{H}_0^{-1}\mathbf{g}_0$$

3. Evaluate the gradient  $\mathbf{g}$  at the new point. If  $\max(\mathbf{g}) < \text{tol}$ , quit. We have converged.

4. Otherwise, update the Hessian using a secant method like SR1.

$$\tilde{\mathbf{H}} = \tilde{\mathbf{H}}_0 + \Delta\tilde{\mathbf{H}}_{\text{SR1}} \quad \Delta\tilde{\mathbf{H}}_{\text{SR1}} = \frac{(\Delta\mathbf{g} - \mathbf{H}_0\Delta\mathbf{x})(\Delta\mathbf{g} - \mathbf{H}_0\Delta\mathbf{x})^t}{(\Delta\mathbf{g} - \mathbf{H}_0\Delta\mathbf{x}) \cdot \Delta\mathbf{x}}$$

5. Set  $\mathbf{x}_0 \leftarrow \mathbf{x}$ ,  $\mathbf{g}_0 \leftarrow \mathbf{g}$ , and  $\mathbf{H}_0 \leftarrow \tilde{\mathbf{H}}$  and return to step 2.