

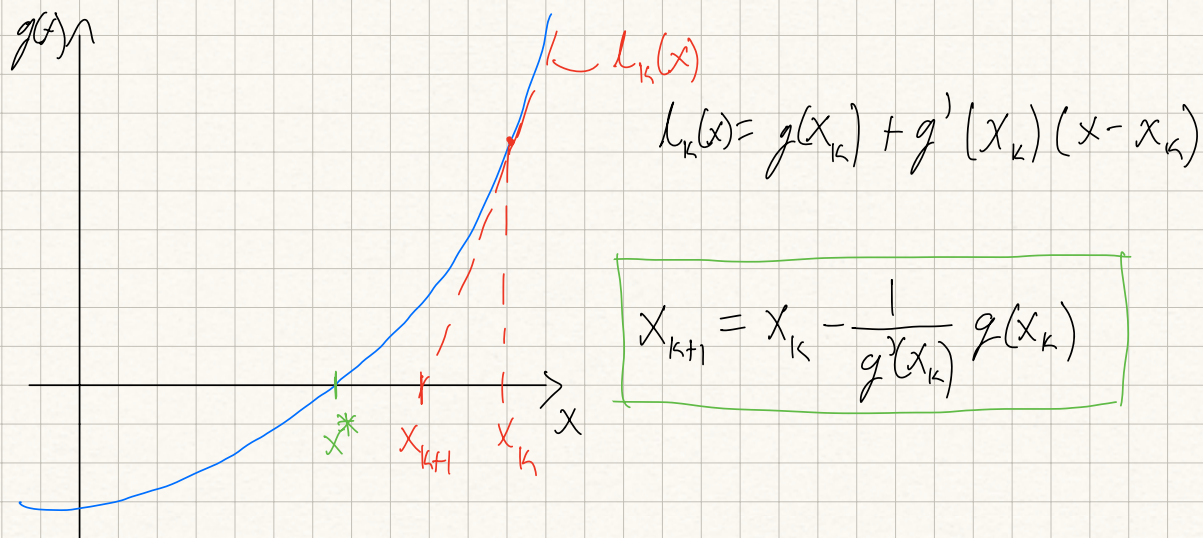
# Lecture 6 - Newton and Quasi-Newton methods

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## Newton-Raphson method

a strictly convex function of one variable.

$g(x) = f'(x)$ , find  $g(x) = 0$



In multivariable:

$$L_k(x) = g_k + J(x_k)(x - x_k)$$

## Newton method for minimizing functions

a) Solve  $\nabla f(x^*) = 0$

$$L_k(x) = \nabla f(x_k) + \nabla^2 f(x_k)(x - x_k) = 0 \Big|_{x = x_{k+1}}$$

$$x_{k+1} = x_k - [\nabla^2 f(x_k)]^{-1} \nabla f(x_k)$$

b) Minimize a quadratic form of  $f$

$$h_k(d) = f(x_k) + \nabla f(x_k)^T d + \frac{1}{2} d^T \nabla^2 f(x_k) d$$

$h_k(d)$  approximates  $f(x_k + d) \quad \forall d \in \mathbb{R}^n$

Select  $x_{k+1} = x_k + d_k$  with

$$d_k = \arg\min h_k(d)$$

$$\nabla h_k(d) = \nabla^2 f(x_k) d + \nabla f(x_k)$$

at  $d_k$  we have

$$\nabla h_k(d_k) = 0 \Rightarrow \nabla^2 f(x_k) d_k + \nabla f(x_k) = 0$$

$$\Rightarrow d_k = -[\nabla^2 f(x_k)]^{-1} \nabla f(x_k)$$

$$\Rightarrow x_{k+1} = x_k + d_k = x_k - [\nabla^2 f(x_k)]^{-1} \nabla f(x_k)$$

The local convergence of the Newton method is quadratic.

Observations about the Newton method:

1) Only good convergence rate locally

2) A costly method computationally;

- have to calculate the gradient

- have to calculate the Hessian



- have to take the inverse of the Hessian

Newton method:

$$x_{k+1} = x_k + d_k, \text{ where}$$

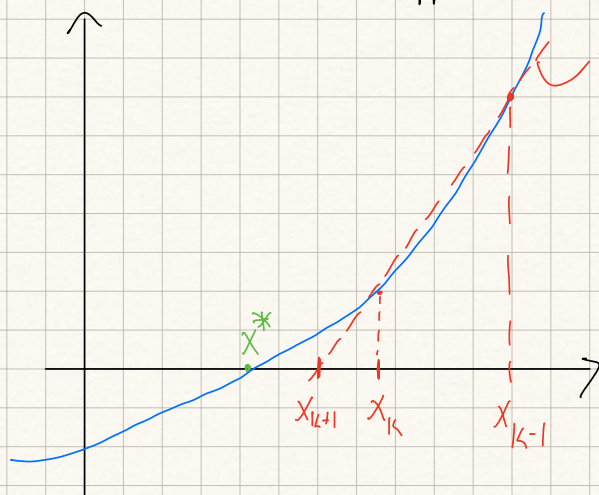
$$d_k = -[\nabla^2 f(x_k)]^{-1} \nabla f(x_k)$$

## Quasi-Newton Method

The Newton method is a tangent method (it uses the tangent of the graph  $g(x)$  at point  $x_k$ ).

Now, we will use a secant method.

We will use an approximation of  $g'(x_k)$



$$S_k = \frac{g(x_{k-1}) - g(x_k)}{x_{k-1} - x_k}$$

$$L_k(x) = g(x_k) + S_k(x - x_k)$$

$$L_k(x) = g(x_k) + g'(x_k)(x - x_k)$$

Approximate  $g'(x_k)$  with  $S_k$ .

$S_k$  is the slope of the secant line connecting  $(g(x_{k-1}), x_{k-1})$  and  $(g(x_k), x_k)$

$$x_{k+1} = x_k - \frac{1}{S_k} g(x_k)$$

Mean value theorem:

$f$  continuous on  $[a, b]$ , differentiable on  $(a, b)$ , there exists a  $a < c < b$  such that

$$f'(c) = \frac{f(b) - f(a)}{b - a}$$

slope of tangent of graph at  $c$  = slope of secant line between  $b$  and  $a$

$$f'(c)(b-a) = f(b) - f(a)$$

in our case, this becomes:

$$\nabla^2 f(c)(x_{k+1} - x_k) = \nabla f(x_{k+1}) - \nabla f(x_k)$$

$$c \in (x_{k+1}, x_k)$$

$$\begin{aligned}\nabla f(x_k) &= \nabla f(x_{k+1}) - \nabla^2 f(c)(x_{k+1} - x_k) \\ &= \nabla f(x_{k+1}) + \nabla^2 f(c)(x_k - x_{k+1})\end{aligned}$$

If  $S_k$  is a good approximation of the Hessian  $\nabla^2 f(x_k)$  we have

$$\nabla f(x_k) = \nabla f(x_{k+1}) + S_k(x_k - x_{k+1})$$

Define  $\gamma_k = \nabla f(x_k) - \nabla f(x_{k+1})$ ,  $\delta_k = x_k - x_{k+1}$  gives

$$\boxed{S_k \delta_k = \gamma_k} \quad \text{Quasi-Newton equation}$$



Usually,  $S_0 = I$