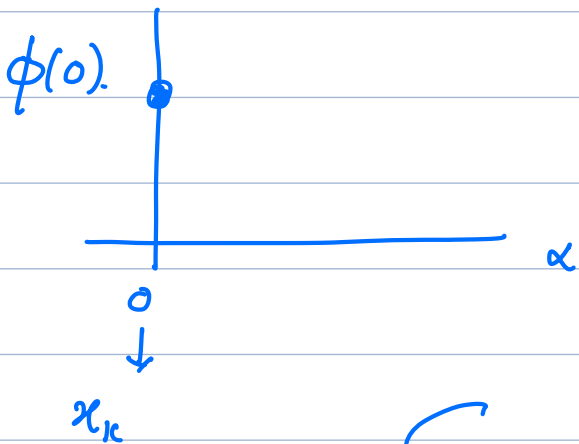


Backtracking line search

repeat until $\left[f(x_k + \alpha p_k) \leq f(x_k) + c \alpha \nabla f_k^T p_k \right]$
 $\alpha \leftarrow \alpha \rho$

Where $0 < \rho < 1$. Typically $\mathcal{L}^{(0)} = 1$.



$$\phi(\alpha) = f(x_k + \alpha p_k)$$

$$\phi'(\alpha) = \nabla f(x_k + \alpha p_k)^T p_k$$

$$\phi'(0) = -\nabla f(x_k)^T P_k$$

3 pieces of data

$$1) \phi(0), 2) \phi'(0), 3) \phi(\alpha^{(0)})$$

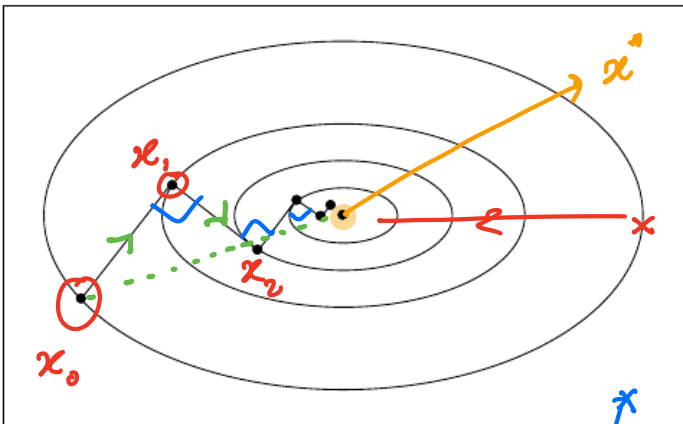
can fit a quadratic fn: $q(x)$

Then? find the minima.

Interpolation to find a poly approx of $\phi(x)$.

Line Search - Analysis.

①



- 1) Zig Zag
- 2) stepsizes are decreasing.

3) angles are 90° betⁿ steps.

for exact line search.

$$S_{k+1} = x_{k+1} - x_k \therefore \text{obsv \#3} \Rightarrow$$

$$S_{k+2}^T S_{k+1} \stackrel{?}{=} 0$$

$$x_{k+1} = x_k + \alpha_k p_k \Rightarrow S_{k+1} = \alpha_k p_k$$

$$= \alpha_k \alpha_{k+1} \begin{pmatrix} p_{k+1}^T & p_k \end{pmatrix}$$

$$\phi(\alpha) = f(x_k + \alpha p_k) \text{ . If line search exact:}$$

$$\Rightarrow \boxed{\phi'(\alpha) = 0} = \nabla f(x_k + \alpha p_k)^T p_k = \nabla f(x_{k+1})^T p_k = 0$$

$$\left. \begin{aligned} p_k &= -\nabla f_k \\ p_{k+1} &= -\nabla f_{k+1} \end{aligned} \right\}$$

$$\Rightarrow S_{k+2}^T S_{k+1} = \alpha_k \alpha_{k+1} \begin{pmatrix} p_{k+1}^T & p_k \end{pmatrix} = 0.$$

② Convergence & Rate.

Yes! $\rightarrow$ It achieves global convergence.
 $\Rightarrow$ Wherever I start $\rightarrow$ I reach a stationary point.

$$a) \cos \theta_k = \frac{(-\nabla f_k^T) p_k}{\|\nabla f_k\| \|p_k\|} \quad \theta_k = 0 \rightarrow \text{gradient descent}$$

$\nearrow$ If descent direction $-90 < \theta_k < 90$

$$\underline{\cos \theta_k \geq 0.}$$

b) Zoutendijk - Thm. $x_{k+1} = x_k + \alpha_k p_k$
↓
descent dir

satisfies Wolfe condns

↳ f is bounded from below

↳ f is continuously differentiable.

↳ the gradient of f is Lipschitz conts.

$$\Rightarrow \|\nabla f(x) - \nabla f(y)\| \leq L \|x - y\|$$

$$\forall x, y \in \mathcal{N}$$

Curvature condns :

$$\Rightarrow \nabla f(x_k + \alpha p_k)^T p_k \geq c_2 \nabla f(x_k)^T p_k$$

Subtr $\nabla f^T p_k$ both sides.

$$\left[\nabla f(x_k + \alpha p_k)^T - \nabla f(x_k)^T \right] p_k \geq (c_2 - 1) \nabla f(x_k)^T p_k. \quad (1)$$

$$(\nabla f_{k+1} - \nabla f_k)^T p_k \leq \|\nabla f_{k+1} - \nabla f_k\| \|p_k\|$$

$$\leq L \alpha_k \|p_k\|^2 \quad (2)$$

$$\Rightarrow \alpha_k \geq \frac{c_2 - 1}{L} \frac{\nabla f_k^T p_k}{\|p_k\|^2} \quad (3)$$

↳ Sufficient decrease:

$$f_{k+1} \leq f_k + c_1 \alpha \nabla f_k^T p_k \quad (4)$$

$$f_{k+1} \leq f_k - \frac{c_1(1-c_2)}{2} \frac{(\nabla f_k^T p_k)^2}{\|p_k\|^2}$$

$$f_{k+1} \leq f_k - c \cos^2 \theta_k \|\nabla f_k\|^2$$

Sum both sides 0 to k

$$\Rightarrow f_{k+1} \leq f_0 - c \sum_{j=0}^k \cos^2 \theta_j \|\nabla f_j\|^2$$

$$c \sum_{j=0}^k \cos^2 \theta_j \|\nabla f_j\|^2 \leq \underbrace{f_0 - f_{k+1}}_{\text{+ve quantity.}} \sim M$$

$$\Rightarrow \sum_{j=0}^k \underbrace{\cos^2 \theta_j \|\nabla f_j\|^2}_{\text{+ve quantity.}} < \infty$$

$$\lim_{k \rightarrow \infty} \cos^2 \theta_k \|\nabla f_k\|^2 \rightarrow 0$$

$$\begin{aligned} f_1 &< f_0 \\ f_2 &< f_1 \end{aligned}$$

Ensure p_k is a legit descent dir

$$\Rightarrow |\cos \theta_k| > \delta$$

$$\Rightarrow \lim_{k \rightarrow \infty} \|\nabla f_k\| = 0$$

$\Rightarrow$ a stationary point.