

LECTURE 2  
LINE SEARCH METHODS

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$$x^{(k+1)} = x^{(k)} + \alpha_k p_k$$

$\swarrow$  step length       $\swarrow$  search direction

Search Direction

(Pure) Newton direction :  $p_k = - \left[ \nabla^2 f(x^{(k)}) \right]^{-1} \nabla f(x^{(k)})$

BFGS direction :  $p_k = - B_k^{-1} \nabla f(x^{(k)})$

It is essential that  $p_k$  is a descent direction, that is

$$\phi'(0) = \nabla f(x^{(k)})^T p_k < 0$$

where

$$\phi: \mathbb{R} \rightarrow \mathbb{R}, \quad \phi(\alpha) := f(x^{(k)} + \alpha p_k).$$

Newton direction is descent if  $\nabla^2 f(x^{(k)})$  is PD, i.e.,

$$\nabla^2 f(x^{(k)}) \text{ is PD} \iff \left[ \nabla^2 f(x^{(k)}) \right]^{-1} \text{ is PD}$$

$$\Rightarrow \nabla f(x^{(k)})^T p_k = -\nabla f(x^{(k)})^T \left[ \nabla^2 f(x^{(k)}) \right]^{-1} \nabla f(x^{(k)}) < 0$$

Similarly BFGS direction is descent if  $B_k$  is PD.

Wolfe Conditions (Selection of Step Length)

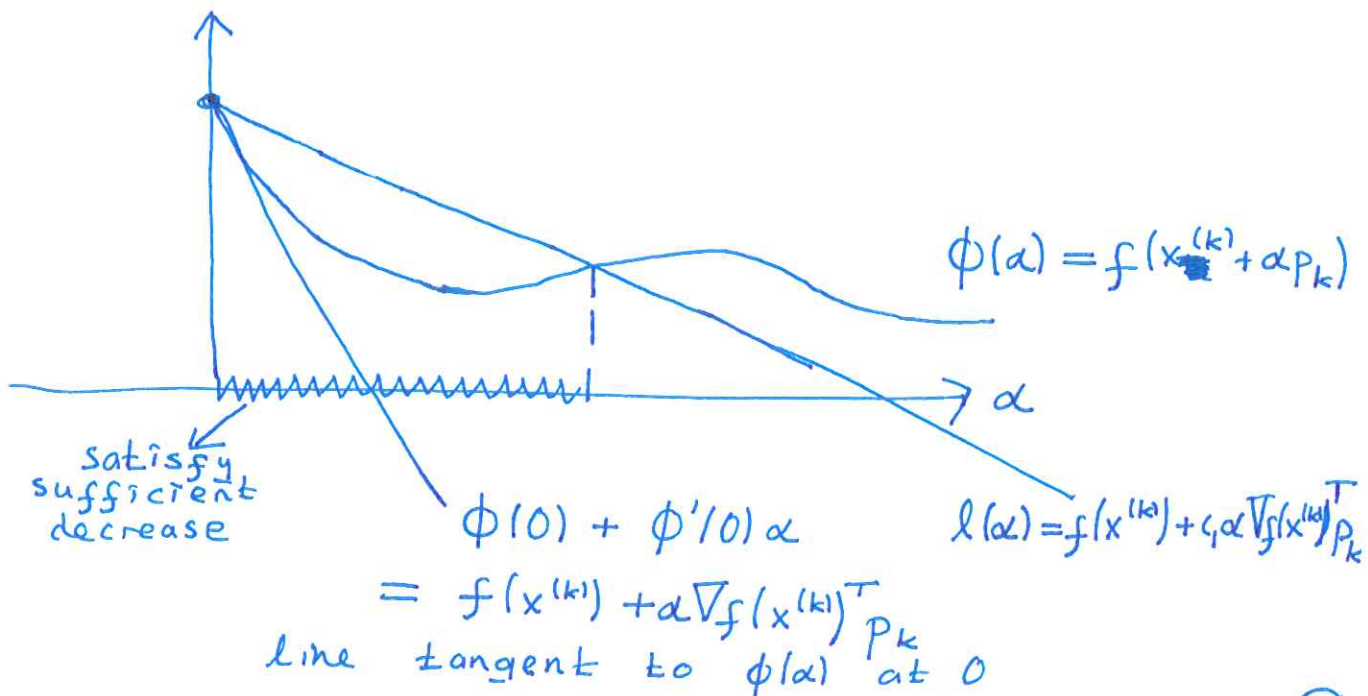
Choose  $\alpha_k$  so that

sufficient decrease  $\leftarrow f(x^{(k)} + \alpha_k p_k) \leq f(x^{(k)}) + c_1 \alpha_k \nabla f(x^{(k)})^T p_k$

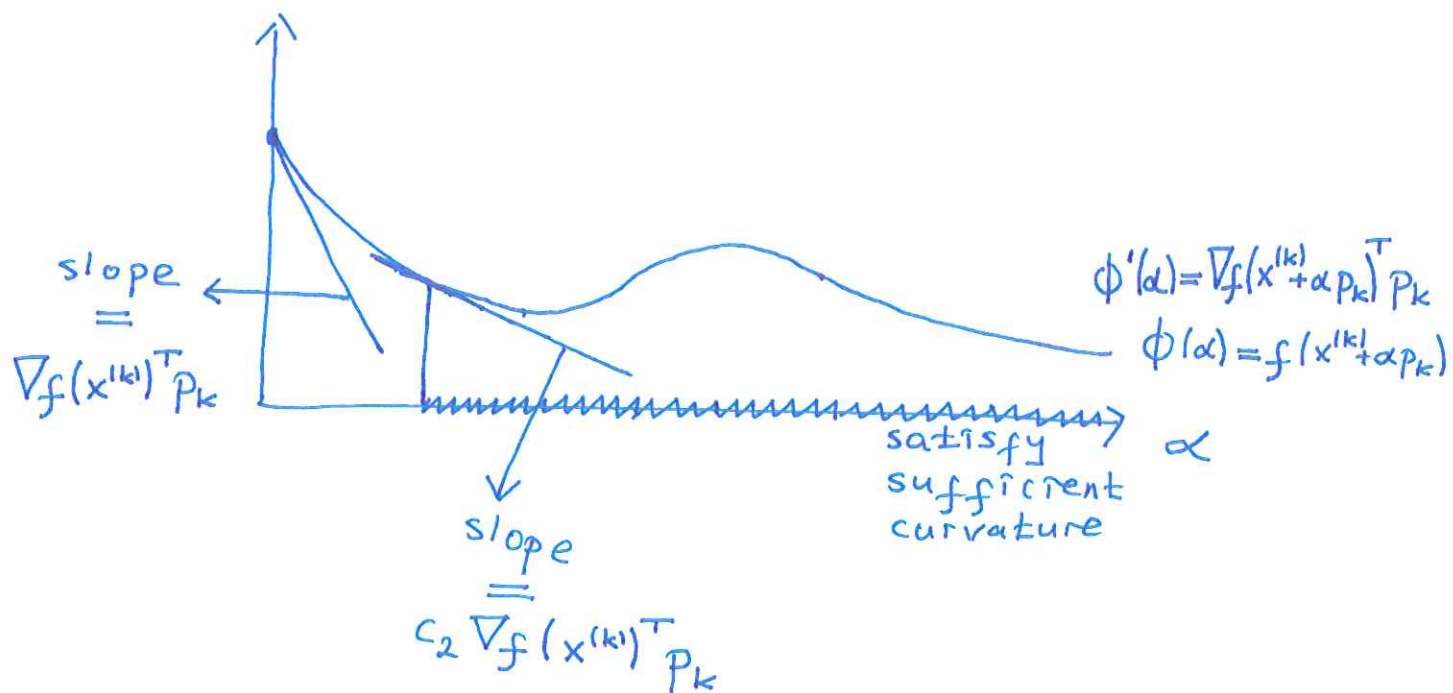
sufficient curvature  $\leftarrow \left[ \nabla f(x^{(k)} + \alpha_k p_k) \right]^T p_k \geq c_2 \nabla f(x^{(k)})^T p_k$

$c_1, c_2$  are given parameters,  $0 < c_1 < c_2 < 1$

Sufficient decrease



# Sufficient curvature



## THM

Suppose  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  is <sup>bounded below,</sup> <sup>twice</sup> continuously differentiable, and  $p_k$  is a descent direction. There exist step-lengths that satisfy Wolfe conditions.

## PROOF

For all  $\alpha \gg 0$  small enough

$$f(x^{(k)} + \alpha p_k) = f(x^{(k)}) + \alpha \nabla f(x^{(k)})^T p_k + O(\alpha^2)$$

$$\leq f(x^{(k)}) + \alpha c_1 \nabla f(x^{(k)})^T p_k := l(\alpha)$$

Furthermore,  $\phi(\alpha)$  is bounded below, whereas  $l(\alpha) \rightarrow -\infty$  as  $\alpha \rightarrow \infty$  meaning

$$\phi(\alpha) \gg l(\alpha)$$

for  $\alpha$  large enough. By continuity there exists  $\alpha \gg 0$  such that  $\phi(\alpha) = l(\alpha)$ . Let  $\underline{\alpha}$  be the smallest such  $\alpha$ .

Sufficient decrease holds on  $[0, \underline{\alpha}]$ .

By the mean value thm there exists

$\underline{\alpha} \in (0, \underline{\alpha})$  such that

$$\begin{aligned} c_1 \nabla f(x^{(k)})^T p_k &= \frac{\phi(\underline{\alpha}) - \phi(0)}{\underline{\alpha}} = \phi'(\underline{\alpha}) \\ &= \nabla f(x^{(k)} + \underline{\alpha} p_k)^T p_k \end{aligned}$$

$$\begin{aligned} \implies & \nabla f(x^{(k)} + \underline{\alpha} p_k)^T p_k > c_2 \nabla f(x^{(k)})^T p_k. \\ \text{Recall} & \\ c_1 < c_2 & \\ \nabla f(x^{(k)})^T p_k & \end{aligned}$$

By continuity of  $\nabla f(x)$  there exists an interval  $I \subseteq [0, \underline{\alpha}]$  containing  $\underline{\alpha}$  such that

$$c_2 \nabla f(x^{(k)})^T p_k < \nabla f(x^{(k)} + \alpha p_k)^T p_k \quad \forall \alpha \in I. \quad \square$$

Strong Wolfe Conditions

$$f(x^{(k)} + \alpha_k p_k) \leq f(x^{(k)}) + c_1 \alpha_k \nabla f(x^{(k)})^T p_k$$

$$\left| \left[ \nabla f(x^{(k)} + \alpha_k p_k) \right]^T p_k \right| \leq c_2 \left| \nabla f(x^{(k)})^T p_k \right|$$



# Global Convergence

## THM (Zoutendijk)

Suppose  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  is bounded below and has Lipschitz continuous gradients, that is there exists  $L > 0$  such that

$$\|\nabla f(x) - \nabla f(y)\|_2 \leq L \|x - y\|_2, \forall x, y.$$

A line search method with descent directions and step-lengths satisfying Wolfe conditions would generate  $\{x^{(k)}\}$  such that

$$\lim_{k \rightarrow \infty} \cos^2 \theta_k \|\nabla f(x^{(k)})\|_2^2 = 0$$

where

$$\theta_k := \arccos \left( \frac{\nabla f(x^{(k)})^T p_k}{\|\nabla f(x^{(k)})\|_2 \|p_k\|_2} \right).$$

## PROOF

Sufficient curvature condition and Lipschitz continuity of  $\nabla f(x)$  would imply

$$\left[ \nabla f(x^{(k+1)}) - \nabla f(x^{(k)}) \right]^T p_k \geq (c_2 - 1) \nabla f(x^{(k)})^T p_k,$$

$$\left[ \nabla f(x^{(k+1)}) - \nabla f(x^{(k)}) \right]^T p_k \leq L \|p_k\|_2^2 \alpha_k$$

$$\Rightarrow \alpha_k \geq \frac{(c_2 - 1) \nabla f(x^{(k)})^T p_k}{L \|p_k\|_2^2}.$$

Exploiting the sufficient decrease condition

$$\begin{aligned} f(x^{(k+1)}) &\leq f(x^{(k)}) + c_1 \alpha_k \nabla f(x^{(k)})^T p_k \\ &\leq f(x^{(k)}) - \frac{c_1(1-c_2) \|\nabla f(x^{(k)})^T p_k\|^2}{L \|p_k\|_2^2} \\ &= f(x^{(k)}) - \frac{c_1(1-c_2)}{L} \cos^2 \theta_k \|\nabla f(x^{(k)})\|_2^2 \end{aligned}$$

$$\Rightarrow \sum_{k=0}^K \cos^2 \theta_k \|\nabla f(x^{(k)})\|_2^2 \leq \frac{L}{c_1(1-c_2)} (f(x^{(0)}) - f(x^{(K)}))$$

Since  $f$  is bounded below, the sum on the left is bounded meaning

$$\lim_{k \rightarrow \infty} \cos^2 \theta_k \|\nabla f(x^{(k)})\|_2^2 = 0.$$

□

Remark

According to Zoutendijk's thm, if there exists  $\delta > 0$  such that

$$|\cos \theta_k| \geq \delta \quad \text{for all } k,$$

then

$$\lim_{k \rightarrow \infty} \nabla f(x^{(k)}) = 0.$$

# Newton's method

suppose

\*  $\nabla^2 f(x^{(k)})$  is PD for  $k = 0, 1, 2, \dots$   
(this also ensures  $p_k$  is a descent direction)

\* there exists  $M$  such that

$$\|\nabla^2 f(x^{(k)})\|_2 \|\left[\nabla^2 f(x^{(k)})\right]^{-1}\|_2 \leq M$$

for  $k = 0, 1, 2, \dots$ , equivalently

$$\frac{\lambda_{\max}(\nabla^2 f(x^{(k)}))}{\lambda_{\min}(\nabla^2 f(x^{(k)}))} \leq M$$

for  $k = 0, 1, 2, \dots$

It follows that

$$|\cos \theta_k| = \frac{|\nabla f(x^{(k)})^T p_k|}{\|\nabla f(x^{(k)})\|_2 \|p_k\|_2}$$

$$= \frac{|\nabla f(x^{(k)})^T \left[\nabla^2 f(x^{(k)})\right]^{-1} \nabla f(x^{(k)})|}{\|\nabla f(x^{(k)})\|_2 \left\| \left[\nabla^2 f(x^{(k)})\right]^{-1} \nabla f(x^{(k)}) \right\|_2}$$

$$\geq \frac{\lambda_{\min} \left( \left[\nabla^2 f(x^{(k)})\right]^{-1} \right) \|\nabla f(x^{(k)})\|_2^2}{\lambda_{\max} \left( \left[\nabla^2 f(x^{(k)})\right]^{-1} \right) \|\nabla f(x^{(k)})\|_2^2}$$

$$= \frac{\lambda_{\min}(\nabla^2 f(x^{(k)}))}{\lambda_{\max}(\nabla^2 f(x^{(k)}))} \geq \frac{1}{M}.$$

Consequently,

$$\lim_{k \rightarrow \infty} \nabla f(x^{(k)}) = 0.$$

Similar holds for BFGS assuming

$B_k$  are PD and  $\|B_k\|_2$   $\|B_k^{-1}\|_2$  remain bounded.