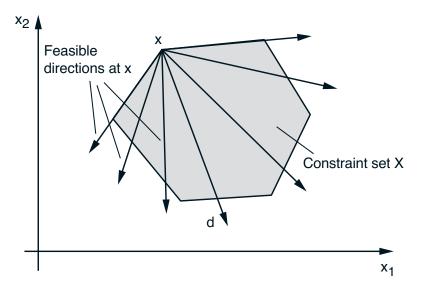
### **6.252 NONLINEAR PROGRAMMING**

### **LECTURE 9: FEASIBLE DIRECTION METHODS**

### LECTURE OUTLINE

- Conditional Gradient Method
- Gradient Projection Methods

A *feasible direction* at an  $x \in X$  is a vector  $d \neq 0$ such that  $x + \alpha d$  is feasible for all suff. small  $\alpha > 0$ 



• Note: the set of feasible directions at x is the set of all  $\alpha(z-x)$  where  $z \in X$ ,  $z \neq x$ , and  $\alpha > 0$ 

### **FEASIBLE DIRECTION METHODS**

• A feasible direction method:

 $x^{k+1} = x^k + \alpha^k d^k,$ 

where  $d^k$ : feasible *descent* direction [ $\nabla f(x^k)'d^k < 0$ ], and  $\alpha^k > 0$  and such that  $x^{k+1} \in X$ .

• Alternative definition:

$$x^{k+1} = x^k + \alpha^k (\overline{x}^k - x^k),$$

where  $\alpha^k \in (0, 1]$  and if  $x^k$  is nonstationary,

$$\overline{x}^k \in X, \qquad \nabla f(x^k)'(\overline{x}^k - x^k) < 0.$$

• Stepsize rules: Limited minimization, Constant  $\alpha^k = 1$ , Armijo:  $\alpha^k = \beta^{m_k}$ , where  $m_k$  is the first nonnegative m for which

$$f(x^k) - f\left(x^k + \beta^m (\overline{x}^k - x^k)\right) \ge -\sigma\beta^m \nabla f(x^k)' (\overline{x}^k - x^k)$$

# **CONVERGENCE ANALYSIS**

• Similar to the one for (unconstrained) gradient methods.

• The direction sequence  $\{d^k\}$  is gradient related to  $\{x^k\}$  if the following property can be shown: For any subsequence  $\{x^k\}_{k \in \mathcal{K}}$  that converges to a nonstationary point, the corresponding subsequence  $\{d^k\}_{k \in \mathcal{K}}$  is bounded and satisfies

 $\limsup_{k \to \infty, \, k \in \mathcal{K}} \nabla f(x^k)' d^k < 0.$ 

Proposition (Stationarity of Limit Points) Let  $\{x^k\}$  be a sequence generated by the feasible direction method  $x^{k+1} = x^k + \alpha^k d^k$ . Assume that:

- $\{d^k\}$  is gradient related
- $\alpha^k$  is chosen by the limited minimization rule or the Armijo rule.

Then every limit point of  $\{x^k\}$  is a stationary point.

• Proof: Nearly identical to the unconstrained case.

#### **CONDITIONAL GRADIENT METHOD**

• 
$$x^{k+1} = x^k + \alpha^k (\overline{x}^k - x^k)$$
, where

$$\overline{x}^k = \arg\min_{x \in X} \nabla f(x^k)'(x - x^k).$$

• We assume that X is compact, so  $\overline{x}^k$  is guaranteed to exist by Weierstrass.

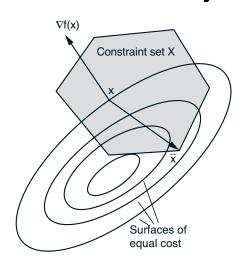
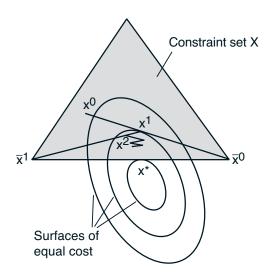


Illustration of the direction of the conditional gradient method.



Operation of the method. Slow (sublinear) convergence.

## **CONVERGENCE OF CONDITIONAL GRADIENT**

• Show that the direction sequence of the conditional gradient method is gradient related, so the generic convergence result applies.

• Suppose that  $\{x^k\}_{k \in K}$  converges to a nonstationary point  $\tilde{x}$ . We must prove that

 $\{x^k - x^k\}_{k \in K} : \text{bounded}, \quad \limsup_{k \to \infty, k \in K} \nabla f(x^k)'(\overline{x}^k - x^k) < 0$ 

- 1st relation: Holds because  $\overline{x}^k \in X$ ,  $x^k \in X$ , and X is assumed compact.
- 2nd relation: Note that by definition of  $\overline{x}^k$ ,

 $\nabla f(x^k)'(\overline{x}^k - x^k) \le \nabla f(x^k)'(x - x^k), \qquad \forall x \in X$ 

Taking limit as  $k \to \infty$ ,  $k \in K$ , and min of the RHS over  $x \in X$ , and using the nonstationarity of  $\tilde{x}$ ,

 $\limsup_{k \to \infty, \, k \in K} \nabla f(x^k)'(\overline{x}^k - x^k) \le \min_{x \in X} \nabla f(\tilde{x})'(x - \tilde{x}) < 0,$ 

thereby proving the 2nd relation.

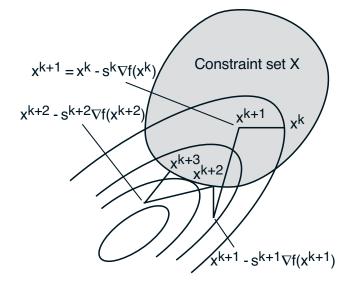
## **GRADIENT PROJECTION METHODS**

 Gradient projection methods determine the feasible direction by using a quadratic cost subproblem. Simplest variant:

$$x^{k+1} = x^k + \alpha^k (\overline{x}^k - x^k)$$

$$\overline{x}^k = \left[x^k - s^k \nabla f(x^k)\right]^+$$

where,  $[\cdot]^+$  denotes projection on the set X,  $\alpha^k \in (0, 1]$  is a stepsize, and  $s^k$  is a positive scalar.



Gradient projection iterations for the case

$$\alpha^k \equiv 1, \quad x^{k+1} \equiv \overline{x}^k$$

If  $\alpha^k < 1$ ,  $x^{k+1}$  is in the line segment connecting  $x^k$  and  $\overline{x}^k$ .

• Stepsize rules for  $\alpha^k$  (assuming  $s^k \equiv s$ ): Limited minimization, Armijo along the feasible direction, constant stepsize. Also, Armijo along the projection arc ( $\alpha^k \equiv 1, s^k$ : variable).

# CONVERGENCE

• If  $\alpha^k$  is chosen by the limited minimization rule or by the Armijo rule along the feasible direction, every limit point of  $\{x^k\}$  is stationary.

• Proof: Show that the direction sequence  $\{\overline{x}^k - x^k\}$  is gradient related. Assume  $\{x^k\}_{k \in K}$  converges to a nonstationary  $\tilde{x}$ . Must prove

$$\{x^k - x^k\}_{k \in K} : \text{bounded}, \quad \limsup_{k \to \infty, k \in K} \nabla f(x^k)'(\overline{x}^k - x^k) < 0$$

1st relation holds because  $\{\|\overline{x}^k - x^k\|\}_{k \in K}$  converges to  $\|[\tilde{x} - s \nabla f(\tilde{x})]^+ - \tilde{x}\|$ . By optimality condition for projections,  $(x^k - s \nabla f(x^k) - \overline{x}^k)'(x - \overline{x}^k) \leq 0$  for all  $x \in X$ . Applying this relation with  $x = x^k$ , and taking limit,

$$\limsup_{k \to \infty, \ k \in K} \nabla f(x^k)'(\overline{x}^k - x^k) \le -\frac{1}{s} \left\| \tilde{x} - \left[ \tilde{x} - s \nabla f(\tilde{x}) \right]^+ \right\|^2 < 0$$

• Similar conclusion for constant stepsize  $\alpha^k = 1$ ,  $s^k = s$  (under a Lipschitz condition on  $\nabla f$ ).

• Similar conclusion for Armijo rule along the projection arc.

#### **CONVERGENCE RATE – VARIANTS**

• Assume  $f(x) = \frac{1}{2}x'Qx - b'x$ , with Q > 0, and a constant stepsize ( $a^k \equiv 1$ ,  $s^k \equiv s$ ). Using the nonexpansiveness of projection

$$\begin{aligned} \left\| x^{k+1} - x^* \right\| &= \left\| \left[ x^k - s \nabla f(x^k) \right]^+ - \left[ x^* - s \nabla f(x^*) \right]^+ \right\| \\ &\leq \left\| \left( x^k - s \nabla f(x^k) \right) - \left( x^* - s \nabla f(x^*) \right) \right\| \\ &= \left\| (I - sQ)(x^k - x^*) \right\| \\ &\leq \max \left\{ |1 - sm|, |1 - sM| \right\} \left\| x^k - x^* \right\| \end{aligned}$$

where m, M: min and max eigenvalues of Q.

• Scaled version:  $x^{k+1} = x^k + \alpha^k (\overline{x}^k - x^k)$ , where

 $\overline{x}^k = \arg\min_{x \in X} \left\{ \nabla f(x^k)'(x - x^k) + \frac{1}{2s^k}(x - x^k)'H^k(x - x^k) \right\},$ and  $H^k > 0$  (involves transformation  $y^k = (H^k)^{1/2}x^k$ . Since the minimum value above is negative when  $x^k$  is nonstationary,  $\nabla f(x^k)'(\overline{x}^k - x^k) < 0$ .

• Newton's method for  $H^k = \nabla^2 f(x^k)$ .

 Variants: Projecting on an expanded constraint set, projecting on a restricted constraint set, combinations with unconstrained methods, etc.