# Duality in Nonlinear Programming

- Given a nonlinear programming problem, there is another nonlinear programming problem closely associated with it
  - primal problem / dual problem
  - the same optimum objective function values under certain convexity assumptions: local convexity, *local duality theory*
  - solve the primal problem indirectly by solving the dual problem

### Local Duality: Equality Constraints

$$\begin{bmatrix} \text{Problem E} \end{bmatrix} \\ \underset{\mathbf{x}}{\text{Minimize}} f(\mathbf{x}) \\ \text{subject to} \\ h_{i}(\mathbf{x}) = 0, \ i = 1, \dots, p \end{bmatrix} \xrightarrow{[\text{Dual Function}]}_{\text{near } \mathbf{v}^{*}, \text{ minimum is taken locally with respect to } \mathbf{x} \text{ near } \mathbf{x}^{*} \\ \begin{array}{c} \phi(\mathbf{v}) = \text{Minimize } L(\mathbf{x}, \mathbf{v}) \left[ = f(\mathbf{x}) + \sum_{i=1}^{p} v_{i} h_{i} \right] \\ \hline Maximize & \phi(\mathbf{v}) \\ Maximize & \phi(\mathbf{v}) \\ \end{array} \\ \end{array}$$

constrained problem in  $x \leftrightarrow$  unconstrained problem in v

$$L(\mathbf{x}, \mathbf{v}) = f(\mathbf{x}) + \sum_{i=1}^{p} v_i h_i = f(\mathbf{x}) + (\mathbf{v} \cdot \mathbf{h})$$
  

$$\mathbf{H}_x(\mathbf{x}, \mathbf{v}) = \frac{\partial^2 L}{\partial \mathbf{x}^2} = \frac{\partial^2 f(\mathbf{x})}{\partial \mathbf{x}^2} + \sum_{i=1}^{p} v_i \frac{\partial^2 h_i}{\partial \mathbf{x}^2}; \ \mathbf{N} = \left[\frac{\partial h_j}{\partial x_i}\right]_{n \times p}$$
  
assumption: 
$$\mathbf{H}_x(\mathbf{x}^*, \mathbf{v}^*) \text{ positive definite} \rightarrow L(\mathbf{x}^*, \mathbf{v}^*) \text{ is locally convex}$$
  

$$\rightarrow \mathbf{x}^* \text{ to be an isolated local minimum of Problem E}$$
  
and a local minimum for the unconstrained problem  $\left[\min_{\mathbf{x}} L(\mathbf{x}, \mathbf{v}^*)\right]$   
vicinity of  $(\mathbf{x}^*, \mathbf{v}^*)$ :  $\frac{\partial L(\mathbf{x}, \mathbf{v})}{\partial \mathbf{x}} = 0 \rightarrow \min_{\mathbf{x}} L(\mathbf{x}, \mathbf{v})$ 

**Optimization Techniques** 

#### Lemmas and Theorem

[Lemma]

The gradient of the dual function 
$$\phi(\mathbf{v})$$
 is given as  $\frac{\partial \phi(\mathbf{v})}{\partial \mathbf{v}} = \mathbf{h}(\mathbf{x}(\mathbf{v}))$ 

 $\rightarrow$  simple to calculatae

The Hessian of the dual function  $\phi(\mathbf{v})$  is given as  $\mathbf{H}_{v} = \frac{\partial^{2} \phi(\mathbf{v})}{\partial \mathbf{v}^{2}} = -\mathbf{N}^{T} [\mathbf{H}_{x}(\mathbf{x})]^{-1} \mathbf{N}$  $\rightarrow$  negative definite

[Local Duality Theorem]

For Problem E, let  $\begin{cases} \mathbf{x}^* \text{ be a local minimum and a regular point} \\ \mathbf{v}^* \text{ be the Lagrange multipliers at } \mathbf{x}^* \\ \mathbf{H}_x(\mathbf{x}^*, \mathbf{v}^*) \text{ be positive definite} \end{cases}$ Then the dual problem [Maximize  $\phi(\mathbf{v})$ ] has a local solution at  $\mathbf{v}^*$  with  $\mathbf{x}^* = \mathbf{x}(\mathbf{v}^*)$ . The maximum value of the dual function is equal to the minimum value of  $f(\mathbf{x})$ ;

that is,  $\phi(\mathbf{v}^*) = f(\mathbf{x}^*)$ 

**Optimization Techniques** 

# Example 5.7

• Consider the following problem in two variables; derive the dual of the problem and solve it:

Minimize 
$$f(x_1, x_2) = -x_1 x_2$$
  
subject to  $h(x_1, x_2) = (x_1 - 3)^2 + x_2^2 = 5$   
 $\rightarrow x_1^* = 4, x_2^* = 2, v^* = 1, f^* = -8$ 

#### Local Duality: Inequality Constraints

$$\begin{bmatrix} \text{Problem P} \end{bmatrix} \\ \underset{\mathbf{x}}{\text{Minimize}} \quad f(\mathbf{x}) \\ \text{subject to} \\ h_i(\mathbf{x}) = 0, \quad i = 1, \dots, p \\ g_j(\mathbf{x}) = 0, \quad j = 1, \dots, m \end{bmatrix} \xrightarrow{[\text{Dual Function}]}_{\mathbf{x}} \text{minimum is taken locally with respect to } \mathbf{x} \text{ near } \mathbf{x}^* \\ \underbrace{\phi(\mathbf{v}, \mathbf{u}) = \text{Minimize}}_{\mathbf{x}} L(\mathbf{x}, \mathbf{v}, \mathbf{u}) \begin{bmatrix} f(\mathbf{x}) + \sum_{i=1}^{p} v_i h_i + \sum_{j=1}^{m} u_j g_j \end{bmatrix}, u_j \ge 0}_{\mathbf{x}} \begin{bmatrix} \text{Dual Problem} \end{bmatrix} \\ \text{Maximize} \quad \phi(\mathbf{v}, \mathbf{u}) \\ u_j \ge 0, \quad j = 1, \dots, m \end{bmatrix}$$

constrained problem in  $\mathbf{x} \leftrightarrow$  unconstrained problem in  $\mathbf{v}, \mathbf{u}$ 

$$L(\mathbf{x}, \mathbf{v}, \mathbf{u}) = f(\mathbf{x}) + \sum_{i=1}^{p} v_i h_i + \sum_{j=1}^{m} u_j g_j = f(\mathbf{x}) + (\mathbf{v} \cdot \mathbf{h}) + (\mathbf{u} \cdot \mathbf{g}), \ u_j \ge 0$$

# Duality Theorem (1)

[Strong Duality Theorem]

For Problem P, let  $\begin{cases} \mathbf{x}^* \text{ be a local minimum and a regular point} \\ \mathbf{v}^*, \mathbf{u}^* \text{ be the Lagrange multipliers at the optimum point } \mathbf{x}^* \\ \mathbf{H}_x(\mathbf{x}^*, \mathbf{v}^*, \mathbf{u}^*) \text{ be positive definite} \end{cases}$ 

Then  $\mathbf{v}^*$  and  $\mathbf{u}^*$  solve the dual problem [Maximize  $\phi(\mathbf{v}, \mathbf{u})$ ] with  $f(\mathbf{x}^*) = \phi(\mathbf{v}^*, \mathbf{u}^*)$  and  $\mathbf{x}^* = \mathbf{x}(\mathbf{v}^*, \mathbf{u}^*)$ 

if  $\mathbf{H}_{x}(\mathbf{x}^{*}, \mathbf{v}^{*}, \mathbf{u}^{*})$  is NOT positive definite  $\rightarrow$  [Weak Duality Theorem] Let  $\mathbf{x}$  be a feasible solution for Problem P and let  $\mathbf{v}$  and  $\mathbf{u}$  be the feasible solution for the dual problem; thus,  $h_{i}(\mathbf{x}) = 0$ , i = 1, ..., p and  $g_{j}(\mathbf{x}) = 0$ , j = 1, ..., mThen  $\phi(\mathbf{v}, \mathbf{u}) \leq f(\mathbf{x})$ 

# Duality Theorem (2)

1. Minimum  $[f(\mathbf{x}) \text{ with } \mathbf{x} \in S] \ge \text{Maximum } [\phi(\mathbf{v}, \mathbf{u}) \text{ with } u_j \ge 0, j = 1, ..., m]$ 2. If  $f(\mathbf{x}^*) = \phi(\mathbf{v}^*, \mathbf{u}^*)$  with  $u_j \ge 0, j = 1, ..., m$  and  $\mathbf{x}^* \in S$ , then  $\mathbf{x}^*$  and  $(\mathbf{v}^*, \mathbf{u}^*)$  solve the primal and dual problem, respectively. 3. If Minimum  $[f(\mathbf{x}) \text{ with } \mathbf{x} \in S] = -\infty$ , then the dial is infeasible, and vice versa (ie, dual is infeasible, the primal is unbounded) 4. If Maximum  $[\phi(\mathbf{v}, \mathbf{u}) \text{ with } u_j \ge 0, j = 1, ..., m] = \infty$ , then the primal problem has no feasible solution, and vice versa (ie, primal is infeasible, the dual is unbounded)

### Saddle Point Theorem

$$\begin{bmatrix} \text{Lemma} \end{bmatrix} \begin{cases} \text{lower bound for primal cost function} \\ \text{upper bound for dual cost function} \end{cases}$$
  
For any  $\mathbf{v}^*$  and  $\mathbf{u}^*$  with  $u_j \ge 0, j = 1, ..., m$   
 $\phi(\mathbf{v}, \mathbf{u}) \le f(\mathbf{x}^*)$ 

#### [Saddle Point Theorem]

For Problem P let all functions be twice continuously differentiable and let  $L(\mathbf{x}, \mathbf{v}, \mathbf{u})$  be defined as  $L(\mathbf{x}, \mathbf{v}, \mathbf{u}) = f(\mathbf{x}) + (\mathbf{v} \cdot \mathbf{h}) + (\mathbf{u} \cdot \mathbf{g}); u_j \ge 0, j = 1,...,m$ Let  $L(\mathbf{x}^*, \mathbf{v}^*, \mathbf{u}^*)$  exist with  $u_j^* \ge 0, j = 1,...,m$ . Also let  $\mathbf{H}_x(\mathbf{x}^*, \mathbf{v}^*, \mathbf{u}^*)$  be positive definite. Then  $\mathbf{x}^*$  satisfying a suitable constraint qualification is a local minumum of Problem P if and only if  $(\mathbf{x}^*, \mathbf{v}^*, \mathbf{u}^*)$  is a saddle point of the Lagrangian; that is,  $L(\mathbf{x}^*, \mathbf{v}, \mathbf{u}) \le L(\mathbf{x}^*, \mathbf{v}^*, \mathbf{u}^*) \le L(\mathbf{x}, \mathbf{v}^*, \mathbf{u}^*)$ for all  $\mathbf{x}$  near  $\mathbf{x}^*$  and all  $(\mathbf{v}, \mathbf{u})$  near  $(\mathbf{v}^*, \mathbf{u}^*)$ , with  $u_j \ge 0, j = 1,...,m$