

Section 15.3

Constrained Optimization:  
Lagrange Multipliers

- Often times in the real world when we would like to optimize something (either minimize or maximize) we are subject to a constraint
- For example, maximizing profit is constrained by a budget
- Minimizing costs is constrained by the amount works must be paid
- Maximizing the time you spend doing homework is constrained by the amount of sleep you need
- And so on...

- Suppose we want to minimize the function

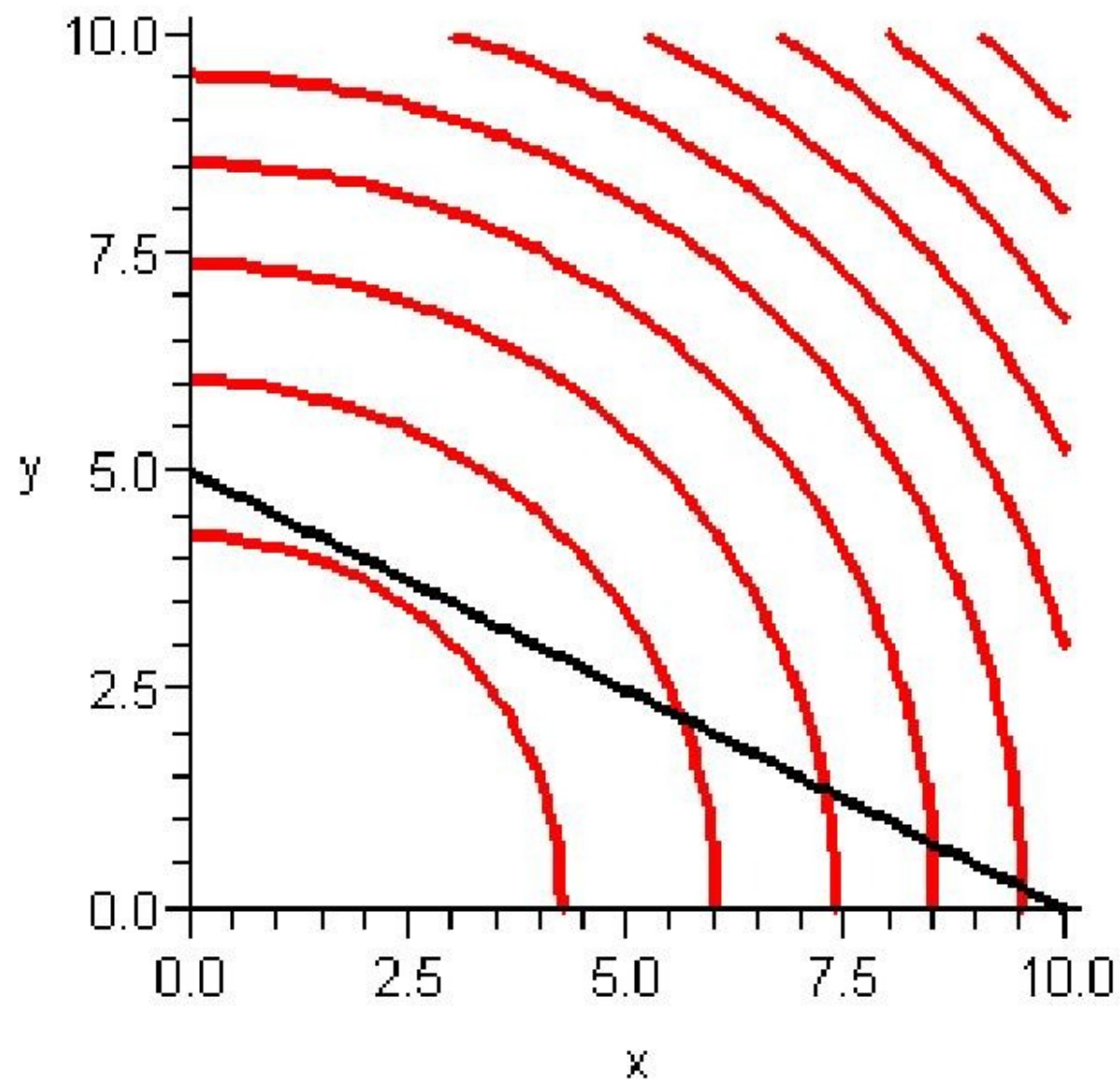
$$f(x, y) = x^2 + y^2$$

subject to the constraint

$$g(x, y) = x + 2y = 10$$

- We refer to  $f$  as the **objective function** and  $g$  as the **constraint equation**
- What we will do is consider graphs of the level curves of  $f$  as well as the constraint equation

Graph of the Level Curves of  $f$   
and the Constraint Equation  $g = 10$



- So we would like to have the level curve that is tangent to the constraint equation
  - This point gives us the smallest value of our function that is also a point on our constraint equation
- We can consider  $g(x,y) = 10$  as a single level curve of  $g(x,y)$ 
  - The value of this is that the gradient of each at their point of intersection should be pointing in the same direction
- So what we want is
 
$$\nabla f(x, y) = \lambda \nabla g(x, y) \text{ and } g(x, y) = 10$$
- Let's work through this problem

# Lagrange Multipliers

- Let  $f$  and  $g$  have continuous partial derivatives. To find the extremum of  $f(x,y)$  subject to the constraint  $g(x,y) = C$  we only need to solve the 3x3 system

$$\begin{cases} f_x(x, y) = \lambda g_x(x, y) \\ f_y(x, y) = \lambda g_y(x, y) \\ g(x, y) = C \end{cases}$$

- $\lambda$  is called the **Lagrange Multiplier**
- If  $g(x,y) = C$  is bounded (and closed) then we are guaranteed both a global min and global max

## Interpretation of Lagrange Multiplier, $\lambda$

- Suppose  $(a,b)$  was an extremum found by using Lagrange Multipliers to optimize  $z = f(x,y)$  subject to  $g(x,y) = C$
- We can consider  $a = a(C)$  and  $b = b(C)$  as functions of  $C$  because as we change  $C$ , our point will change as well
- Now  $z_0 = f(a,b)$  where  $(a,b)$  is a solution to

$$\begin{cases} f_x(x, y) = \lambda g_x(x, y) \\ f_y(x, y) = \lambda g_y(x, y) \end{cases}$$

- Since  $a = a(C)$  and  $b = b(C)$ ,  $z_0 = f(a,b)$  is a function of  $C$

$$\frac{dz_0}{dC} = f_x(a,b) \frac{da}{dC} + f_y(a,b) \frac{db}{dC}$$

$$= \lambda g_x(a,b) \frac{da}{dC} + \lambda g_y(a,b) \frac{db}{dC}$$

$$= \lambda \left[ g_x(a,b) \frac{da}{dC} + g_y(a,b) \frac{db}{dC} \right]$$

$$= \lambda \left[ \frac{dg}{dC}(a,b) \right] = \lambda \frac{dC}{dC} = \lambda \quad \text{b/c } g(a,b) = C$$



- So we have  $\frac{dz_0}{dC} = \lambda$
- Thus  $\lambda$  is the change in the local extreme value of  $f$  per unit increase in  $C$

# Examples

- Optimize the following functions with the given constraint

$$f(x, y) = 3x - 2y, \quad x^2 + 2y^2 = 44$$

$$g(x, y) = x^2 - xy + y^2, \quad x^2 + y^2 = 1$$

- We will look at both of these in maple