

## 14.8: Lagrange Multipliers

- Can be used to help solve optimization problems that have constraints

### Orthogonal Gradient Theorem

Suppose  $f(x, y, z)$  is differentiable in region whose interior contains smooth curve:

$$C : \vec{r}(t) = x(t)\mathbf{i} + y(t)\mathbf{j} + z(t)\mathbf{k}$$

If  $P_0$  is a point on  $C$  where  $f$  has a local extremum relative to its values on  $C$ ,  $\nabla f$  is orthogonal to  $C$  at  $P_0$ .

#### Corollary

At the points on a smooth curve  $\vec{r}(t) = x(t)\mathbf{i} + y(t)\mathbf{j}$  where a differentiable function  $f(x, y)$  takes on its local extrema relative to its values on the curve,  $\nabla f \cdot \vec{r}' = 0$ .

### Method

Suppose  $f(x, y, z)$  and  $g(x, y, z)$  are differentiable and  $\nabla g \neq 0$  when  $g(x, y, z) = 0$ . To find local extremum of  $f$  subject to  $g(x, y, z) = 0$ , find  $x, y, z, \lambda$  satisfying:

$$\begin{aligned}\nabla f &= \lambda \nabla g \\ g(x, y, z) &= 0\end{aligned}$$

### Example

Maximize  $xy$  on ellipse  $4x^2 + 9y^2 = 36$ .

$$\begin{aligned}f(x, y) &= xy \\ \nabla f(x, y) &= y\mathbf{i} + x\mathbf{j}\end{aligned}$$

$$\begin{aligned}g(x, y) &= 4x^2 + 9y^2 - 36 \\ \nabla g(x, y) &= 8x\mathbf{i} + 18y\mathbf{j}\end{aligned}$$

Equations formed:

$$\begin{aligned}y &= \lambda(8x) \\ x &= \lambda(18y) \\ 4x^2 + 9y^2 - 36 &= 0\end{aligned}$$

We get values for  $x$  and  $y$ . This gives us points we can use to maximize  $xy$ .  
 $\lambda$  is unused.

## 14.9: Taylor's Formula for $f(x, y)$

### Taylor Polynomial (recap)

If function  $f$  has  $n$  derivatives at point where  $x = a$ , then the  $n$ th Taylor Polynomial for  $f$  at  $a$  is:

$$P_n(x) = \sum_{k=0}^n \frac{f^{(k)}(a)(x-a)^k}{k!}$$

### The theorem

If  $f$  has  $n+1$  derivatives on an open interval containing  $a$ , then for every  $x$  in that open interval, we have:

$$f(x) = P_n(x) + \frac{f^{(n+1)}(c)}{(n+1)!} (x-a)^{n+1}$$

for some (estimated) value  $c$  between  $a$  and  $x$  that maximizes that term.

The absolute value of new term is called the error when using  $P_n(x)$  to approximate  $f(x)$ .

$$\text{error} = |f(x) - P_n(x)| = \frac{|f^{(n+1)}(c)|}{(n+1)!} |x-a|^{n+1}$$

Give an error estimate for the approximation of  $\cos(2x)$  by  $P_{10}(x)$  for an arbitrary  $x$  between 0 and  $\pi/4$  centered at  $x=0$ .

$$\text{error} \leq \frac{|f^{(n+1)}(c)|}{(n+1)!} (x-a)^{n+1} \quad \leftarrow n+1=11$$

$$f'(x) = -2 \sin(2x)$$

$$f''(x) = -4 \cos(2x)$$

$$f'''(x) = 8 \sin(2x)$$

$$\leftarrow 2^3$$

$$|f^{(11)}(x)| = 2^{11} \cdot |\sin(2x) \text{ or } \cos(2x)|$$

$$f^{(11)}(c) \leq 2048 (1)$$

$$\text{error} \leq \frac{2048}{11!} (\pi/4)^{11}$$

(I believe this was done in BC)

## Two Variables

Suppose  $f(x, y)$  and its partials thru order  $n + 1$  are continuous throughout open rectangular region  $R$  centered around  $(a, b)$ . Then, throughout  $R$ :

$$f(a + h, b + k) = \sum_{i=0}^{n+1} \frac{1}{i!} \left( h \frac{\partial}{\partial x} + k \frac{\partial}{\partial y} \right)^i f \Big|_{(a,b)}$$

Error term is the last one, last is also an approximate error term.

## 14.10: Partial Derivatives w/ Constraints

Steps

1. Decide which variables are dependent & independent
2. Eliminate the other dependent variables
3. Differentiate and solve

### Example

If  $w = x^2 + y - z + \sin(t)$  and  $x + y = t$ , find  $\left( \frac{\partial w}{\partial y} \right)_{z,t}$

(notation designates that  $z, t$  are independent)

$$\begin{aligned} x &= t - y \\ w &= (t - y)^2 + y - z + \sin(t) \\ \frac{\partial w}{\partial y} &= -2(t - y) + 1 \end{aligned}$$

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