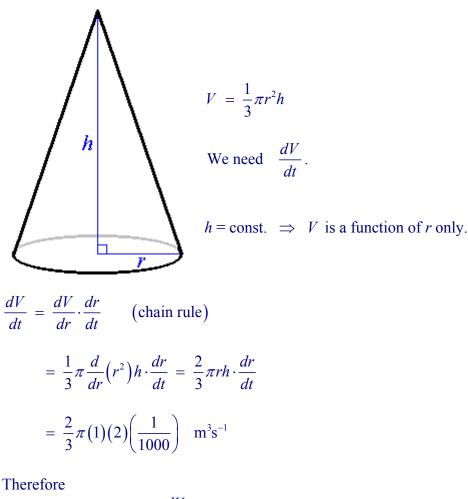
2.1 <u>Partial Derivatives</u>

Example 2.1.1

At a particular instant, a cone has a height of h = 2 m and a base radius of r = 1 m. The base radius is increasing at a rate of 1 mm/s. The height is constant. How fast is the volume V increasing at this time?



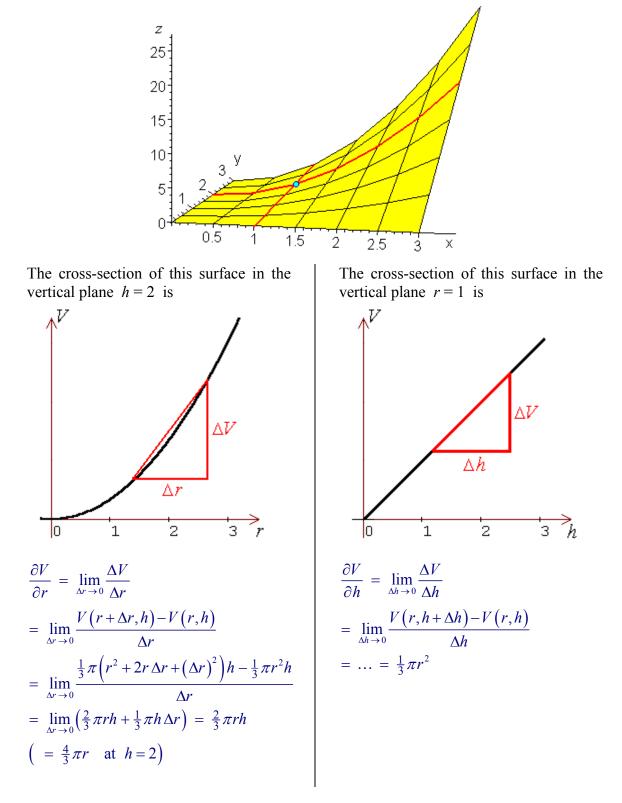
 $\frac{dV}{dt} = \frac{\pi}{750} \text{ m}^3 \text{s}^{-1} \approx 0.00419 \text{ m}^3 \text{s}^{-1}$

But if $\frac{dr}{dt} = 1 \text{ mm s}^{-1}$ and $\frac{dh}{dt} = -2 \text{ mm s}^{-1}$, how do we find $\frac{dV}{dt}$? We shall return to this question later.

Page 2-02

Graph of V against r and h:

Plotting z = V (where $V = \pi r^2 h/3$) against both x = r and y = h yields



At each point on the surface, these two tangent lines define a tangent plane.

V is a function of r and h, each of which in turn is a function of t only. In this case, the chain rule becomes

| dV _ | ∂V | dr | | ∂V | dh |
|--------|--------------|----|---|--------------|----|
| dt – | ∂r | dt | Ŧ | ∂h | dt |

In example 2.1.1,

$$r = 1, h = 2, \frac{\partial V}{\partial r} = \frac{2}{3}\pi rh = \frac{4}{3}\pi, \quad \frac{\partial V}{\partial h} = \frac{1}{3}\pi r^{2} = \frac{1}{3}\pi,$$
$$\frac{dr}{dt} = \frac{1}{1000}, \quad \frac{dh}{dt} = \frac{-2}{1000} \implies \frac{dV}{dt} = \frac{4\pi}{3} \times \frac{1}{1000} - \frac{\pi}{3} \times \frac{2}{1000} = \frac{\pi}{1500} \,\mathrm{m}^{3}\mathrm{s}^{-1}$$

Alternative notations:

$$\frac{\partial V}{\partial r} = V_r = D_r V = \frac{\partial}{\partial r} (V(r,h))$$

If
$$w = f(x, y, z)$$
, then $\frac{\partial w}{\partial y} = \lim_{\Delta y \to 0} \left(\frac{f(x, y + \Delta y, z) - f(x, y, z)}{\Delta y} \right)$, etc.

A Maple worksheet, used to generate the graph of $V(r,h) = \frac{1}{3}\pi r^2 h$, is available at "http://www.engr.mun.ca/~ggeorge/2422/programs/conevolume.mws".

Open this worksheet in Maple and click on the graph.

Then, by dragging the mouse (with left button down), one can change the direction of view of the graph as one wishes. Other features of the graph may be changed upon opening a menu with a right mouse click on the graph or by using the main menu at the top of the Maple window.

Example 2.1.2

$$f(x, y, z) = \sqrt{x^2 + y^2 + z^2}.$$

Find $f_z(0, 3, 4)$
[the first partial derivative of f with respect to z, evaluated at the point (0, 3, 4)].

$$\frac{\partial f}{\partial z} = \frac{1}{2} \left(x^2 + y^2 + z^2 \right)^{-1/2} \cdot \frac{\partial}{\partial z} \left(x^2 + y^2 + z^2 \right)$$
$$= \frac{1}{2f} \cdot \left(0 + 0 + 2z \right) = \frac{z}{f}$$

$$\Rightarrow f_z(0,3,4) = \frac{4}{\sqrt{0+9+16}} = \frac{4}{5}$$

OR

$$f^2 = x^2 + y^2 + z^2$$

Using implicit differentiation,

$$2f\frac{\partial f}{\partial z} = 0 + 0 + 2z$$

$$\Rightarrow f_z = \frac{z}{f} = \frac{4}{5}, \quad (as before)$$

In this example,

f = (distance of the point (x, y, z) from the origin).

Example 2.1.3

$$u = x^{y/z}$$

Find the three first partial derivatives, u_x , u_y , u_z .

$$\frac{d}{dx}(x^{n}) = n x^{n-1} \implies \frac{\partial}{\partial x}(x^{y/z}) = \frac{y}{z} \cdot x^{\left(\frac{y}{z}-1\right)}$$
$$\therefore \quad u_{x} = \frac{y}{z} x^{(y-z)/z} = \frac{yu}{xz}$$

$$u = x^{y/z} = \exp\left(\frac{y}{z}\ln x\right)$$
$$\Rightarrow \quad u_y = \frac{1}{z}(\ln x)\exp\left(\frac{y}{z}\ln x\right) = \frac{u}{z}\ln x \quad \Rightarrow$$
$$u_y = \frac{1}{z}x^{y/z}\ln x$$

OR

$$\ln u = \frac{y}{z} \ln x \implies \frac{\partial}{\partial y} (\ln u) = \frac{\partial}{\partial y} \left(\frac{y}{z} \ln x \right)$$
$$\Rightarrow \frac{1}{u} \cdot u_y = \frac{1}{z} \cdot \ln x \implies u_y = \frac{u}{z} \cdot \ln x$$

$$\ln u = yz^{-1} \ln x \qquad \Rightarrow \quad \frac{1}{u} \cdot u_z = y(-1z^{-2}) \ln x$$
$$\Rightarrow \quad u_z = -\frac{uy}{z^2} \ln x = -\frac{y}{z^2} x^{y/z} \ln x$$

2.2 <u>Higher Partial Derivatives</u>

$$u_{x} = \frac{\partial u}{\partial x}$$

$$\frac{\partial}{\partial x} \left(\frac{\partial u}{\partial x} \right) = \frac{\partial^{2} u}{\partial x^{2}} = u_{xx}$$

$$\frac{\partial}{\partial y} \left(\frac{\partial u}{\partial x} \right) = \frac{\partial^{2} u}{\partial y \partial x} = u_{xy}$$

$$\frac{\partial}{\partial z} \left(\frac{\partial^{2} u}{\partial y \partial x} \right) = \frac{\partial^{3} u}{\partial z \partial y \partial x} = u_{xyz}$$

$$\frac{\partial}{\partial y} \left(\frac{\partial^{2} u}{\partial y \partial x} \right) = \frac{\partial^{3} u}{\partial y^{2} \partial x} = u_{xyy}$$

etc.

Example 2.2.1

$$u = x e^{-t} \sin y$$

Find the second partial derivatives u_{xy} , u_{yx} , u_{xx} and the third partial derivative u_{ttt} .

$$u_{x} = e^{-t} \sin y$$

$$u_{y} = x e^{-t} \cos y$$

$$u_{y} = x e^{-t} \cos y$$

$$u_{t} = -x e^{-t} \sin y = -u$$

$$u_{xy} = \frac{\partial}{\partial y}(u_{x}) = \frac{e^{-t} \cos y}{2}$$

$$u_{yx} = \frac{\partial}{\partial x}(u_{y}) = \frac{e^{-t} \cos y}{2} = u_{xy}$$

$$u_{xx} = 0$$

$$u_{t} = -u$$

$$\Rightarrow u_{tt} = (-u)_{t} = +u$$

$$\Rightarrow u_{tt} = (+u)_{t} = -u$$
Therefore
$$u_{tt} = \frac{-x e^{-t} \sin y}{2}$$

Clairaut's Theorem

If, on a disk *D* containing the point (a, b), *f* is defined and both of the partial derivatives f_{xy} and f_{yx} are continuous, (which is the case for most functions of interest), then

$$f_{xy}(a,b) = f_{yx}(a,b)$$

that is, the order of differentiation doesn't matter.

One of the most important partial differential equations involving second partial derivatives is **Laplace's equation**, which arises naturally in many applications, including electrostatics, fluid flow and heat conduction:

$$\frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} = 0$$

or its equivalent in \mathbb{R}^3 :

$$\frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} + \frac{\partial^2 u}{\partial z^2} = 0$$

Example 2.2.2

Does $u = \ln \sqrt{x^2 + y^2}$ satisfy Laplace's equation?

$$u = \frac{1}{2} \ln \left(x^{2} + y^{2} \right)$$

$$\Rightarrow u_{x} = \frac{\frac{1}{2}(2x)}{x^{2} + y^{2}} = \frac{x}{x^{2} + y^{2}}$$

$$\Rightarrow u_{xx} = \frac{1(x^{2} + y^{2}) - x(2x)}{(x^{2} + y^{2})^{2}} = \frac{y^{2} - x^{2}}{(x^{2} + y^{2})^{2}}$$

u is symmetric with respect to x, y.

Therefore, to find u_{yy} , interchange x, y in u_{xx} .

$$\Rightarrow u_{yy} = \frac{x^2 - y^2}{\left(y^2 + x^2\right)^2} = -u_{xx}$$

 $\Rightarrow u_{xx} + u_{yy} = 0$ Therefore *u* does satisfy Laplace's equation (except at (0,0)).

2.3 <u>Differentials</u>

In \mathbb{R}^2 , let a curve have the Cartesian equation y = f(x).

The small change in y, (Δy) , caused by travelling along the curve for a small horizontal distance Δx , may be approximated by the change dy that is caused by travelling for the same horizontal distance Δx along the tangent line instead.

The exact form is $\Delta y = f(x + \Delta x) - f(x)$.

The approximation to Δy is

 $\Delta y \approx dy = f'(x) dx$

where the differential dx has been replaced by the increment Δx . The approximation improves as Δx decreases towards zero.

Stepping up one dimension, let a surface have the Cartesian equation z = f(x, y). The change in the dependent variable z caused by small changes in the independent variables x and y has the exact value

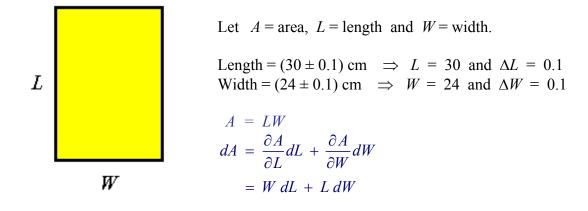
$$\Delta z = f(x + \Delta x, y + \Delta y) - f(x, y) .$$

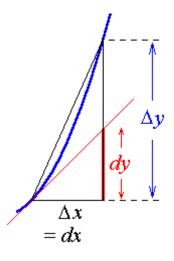
The approximation to Δz is

$$\Delta z \approx dz = \frac{\partial f}{\partial x} dx + \frac{\partial f}{\partial y} dy$$

Example 2.3.1

A rectangle has quoted dimensions of 30 cm for length and 24 cm for width. However, there may be an error of up to 1 mm in the measurement of each dimension. Estimate the maximum error in the calculated area of the rectangle.





Example 2.3.1 (continued)

Let dL = dW = 0.1, then

 $\max(\text{error}) = \Delta A \approx dA$

$$= 24 \times 0.1 + 30 \times 0.1$$
$$= 5.4 \text{ cm}^2.$$

Compare this to a direct calculation:

 $\max(\text{error}) = \max\{ (A_{\max} - A), (A - A_{\min}) \}$ $A - A_{\min} = 30 \times 24 - (30 - 0.1)(24 - 0.1)$ $= 5.39 \text{ cm}^2.$ $A_{\max} - A = (30 + 0.1)(24 + 0.1) - 30 \times 24$

$$= 5.41 \text{ cm}^2$$
.

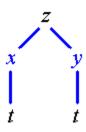
Therefore

 $\max(\text{error}) = \frac{5.41 \text{ cm}^2}{5.41 \text{ cm}^2}.$ Relative error: $\max(\text{error in } L) = L/300$ $\max(\text{error in } W) = W/240$ A = LW $\Rightarrow \quad dA = W \, dL + L \, dW$ $\Rightarrow \quad \frac{dA}{A} = \frac{W \, dL}{LW} + \frac{L \, dW}{LW}$ $\Rightarrow \quad \frac{dA}{A} = \frac{dL}{L} + \frac{dW}{W} = \frac{1}{300} + \frac{1}{240} = \frac{4+5}{1200} = \frac{3}{400}$ $\therefore \quad \frac{\Delta A}{A} \approx \frac{dA}{A} = \frac{0.75\%}{4}$ and 0.75% of $A = 720 \text{ cm}^2$ is 5.4 cm².

Chain Rule

z = f(x, y). If x and y are both functions of t only, then, by the chain rule,

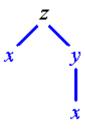
| dz _ | $\partial z dx$ | $\partial z dy$ |
|--------|------------------|------------------|
| dt | $\partial x dt$ | $\partial y dt$ |



If z = f(x, y) and y in turn is a function of x only, then replace t by x in the formula above.

 $\frac{dx}{dx} = 1$ and

$$\frac{dz}{dx} = \frac{\partial z}{\partial x} + \frac{\partial z}{\partial y}\frac{dy}{dx}$$

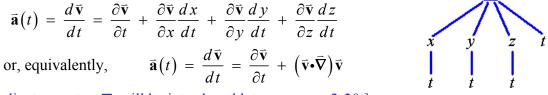


Note the distinction between the total derivative $\frac{dz}{dx}$ and the partial derivative $\frac{\partial z}{\partial x}$.

Example 2.3.2

In the study of fluid dynamics, one approach is to follow the motion of a point in the fluid. In that approach, the velocity vector is a function of both time and position, while position, in turn, is a function of time. $\mathbf{v} = \mathbf{v}(\mathbf{r}, t)$ and $\mathbf{r} = \mathbf{r}(t)$.

The acceleration vector is then obtained through differentiation following the motion of the fluid:



[The gradient operator, ∇ , will be introduced later, on page 2-20.]

Further analysis of an ideal fluid of density ρ at pressure p subjected to a force field F leads to Euler's equation of motion

$$\frac{\partial \vec{\mathbf{v}}}{\partial t} + \left(\vec{\mathbf{v}} \cdot \vec{\nabla} \right) \vec{\mathbf{v}} = \vec{\mathbf{F}} - \frac{\vec{\nabla} p}{\rho}$$

This application of partial differentiation will be explored in some disciplines in a later semester. As a simple example here, suppose that $\bar{\mathbf{v}} = \langle e^{-x}, 1, 10(1-t) \rangle$, then find the acceleration vector.

First note that the velocity vector is the derivative of the displacement vector, so that

$$\bar{\mathbf{v}} = \frac{d\bar{\mathbf{r}}}{dt} = \left\langle \frac{dx}{dt}, \frac{dy}{dt}, \frac{dz}{dt} \right\rangle = \left\langle e^{-x}, 1, 10(1-t) \right\rangle$$

$$\frac{\partial \bar{\mathbf{v}}}{\partial t} = \left\langle 0, 0, -10 \right\rangle, \qquad \frac{\partial \bar{\mathbf{v}}}{\partial x} = \left\langle -e^{-x}, 0, 0 \right\rangle$$

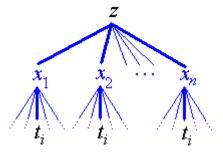
$$\frac{\partial \bar{\mathbf{v}}}{\partial y} = \frac{\partial \bar{\mathbf{v}}}{\partial z} = \left\langle 0, 0, 0 \right\rangle$$

$$\bar{\mathbf{a}}(t) = \frac{d\bar{\mathbf{v}}}{dt} = \frac{\partial \bar{\mathbf{v}}}{\partial t} + \frac{\partial \bar{\mathbf{v}}}{\partial x} \frac{dx}{dt} + \frac{\partial \bar{\mathbf{v}}}{\partial y} \frac{dy}{dt} + \frac{\partial \bar{\mathbf{v}}}{\partial z} \frac{dz}{dt}$$

$$\Rightarrow \frac{d\bar{\mathbf{v}}}{dt} = \left\langle 0, 0, -10 \right\rangle + \left\langle -e^{-x}, 0, 0 \right\rangle e^{-x} + \bar{\mathbf{0}} + \bar{\mathbf{0}} = \left\langle -e^{-2x}, 0, -10 \right\rangle$$
[One can show that $x(t) = \ln |t+c|$, where c is a constant.]

Generalized Chain Rule

Let z be a function of n variables $\{x_1, x_2, ..., x_n\}$, each of which, in turn, is a function of m variables $\{t_1, t_2, ..., t_m\}$, so that $z = f(x_1(t_1, t_2, ..., t_m), x_2(t_1, t_2, ..., t_m), ...$ $\dots, x_n(t_1, t_2, ..., t_m)$). To find $\frac{\partial z}{\partial t_i}$,

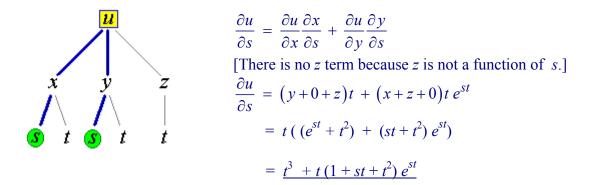


trace all paths that start at z and end at t_i , via all of the $\{x_i\}$ variables.

$$dz = \sum_{j=1}^{n} \frac{\partial z}{\partial x_j} dx_j$$
 and $\frac{\partial z}{\partial t_i} = \sum_{j=1}^{n} \frac{\partial z}{\partial x_j} \frac{\partial x_j}{\partial t_i}$ $(i = 1, 2, \dots, m)$

Example 2.3.3:

u = xy + yz + zx, x = st, $y = e^{st}$ and $z = t^2$. Find u_s in terms of s and t only. Find the value of u_s when s = 0 and t = 1.



This derivative could also be found directly by replacing x, y and z by the respective functions of s and t before differentiating u:

$$u = (st) e^{st} + (e^{st}) t^{2} + (t^{2})st$$

$$\Rightarrow \quad u_{s} = (st^{2} + t) e^{st} + (e^{st}) t^{3} + t^{3} = \underline{t^{3} + t(1 + st + t^{2})} e^{st}$$

$$u_{s}(0, 1) = 1 + 1(1 + 0 + 1)e^{0} = 1 + 2 = \underline{3}.$$

Differentials

Page 2-13

Implicit functions:

If z is defined implicitly as a function of x and y by F(x, y, z) = c, then $dF = F_x dx + F_y dy + F_z dz = 0$ $\Rightarrow dz = -\frac{1}{F_z} (F_x dx + F_y dy)$ provided $F_z = \frac{\partial F}{\partial z} \neq 0$ Example 2.3.4:

Find the change in z when x and y both increase by 0.2 from the point (1, 2, 2) on the sphere $x^2 + y^2 + z^2 = 9$.

$$F = x^{2} + y^{2} + z^{2} = 9 \implies dz = -\frac{1}{2z} (2x \, dx + 2y \, dy) = -\frac{1}{z} (x \, dx + y \, dy)$$

Note:

If y is a function of x only, then

$$\frac{dz}{dx} = \frac{\partial z}{\partial x} + \frac{\partial z}{\partial y}\frac{dy}{dx} = -\frac{x}{z} - \frac{y}{z}\frac{dy}{dx}$$

However, if x and y are independent of each other, then $\frac{dz}{dx}$ is ill-defined. The only derivative of z with respect to x that we can then define is $\frac{\partial z}{\partial x}$.

$$x^2 + y^2 + z^2 = 9 \implies 2x + 0 + 2z \frac{\partial z}{\partial x} = 0 \implies \frac{\partial z}{\partial x} = -\frac{x}{z}$$
. [End note]

Solution – Approximate motion on the sphere by motion on the tangent plane:

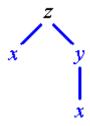
$$x = 1$$
, $y = z = 2$, $dx \approx \Delta x = 0.2$, $dy \approx \Delta y = 0.2$.

$$\Rightarrow \Delta z \approx dz = -\frac{1}{2} (1 \times 0.2 + 2 \times 0.2) = -0.3$$

Therefore z decreases by [approximately] 0.3

Exact: $z_{old} = \sqrt{9 - 1^2 - 2^2} = \sqrt{4} = 2$ $z_{new} = \sqrt{9 - (1.2)^2 - (2.2)^2} = \sqrt{2.72} \approx 1.649$

Therefore $\Delta z = -0.3507...$



Curves of Intersection

Example 2.3.5:

Find both partial derivatives with respect to z on the curve of intersection of the sphere centre the origin, radius 5, and the circular cylinder, central axis on the y-axis, radius 3.

Sphere: $f = x^2 + y^2 + z^2 = 25$

Cylinder:

 \Rightarrow

 $df = 2x \, dx + 2y \, dy + 2z \, dz = 0$

 $g = x^2 + z^2 = 9$

and

 $dg = 2x \, dx \qquad + 2z \, dz = 0$

which leads to the linear system

$$\begin{bmatrix} x & y \\ x & 0 \end{bmatrix} \cdot \begin{bmatrix} dx \\ dy \end{bmatrix} = -\begin{bmatrix} 1 \\ 1 \end{bmatrix} z \, dz$$
$$\Rightarrow \begin{bmatrix} dx \\ dy \end{bmatrix} = -\begin{bmatrix} x & y \\ x & 0 \end{bmatrix}^{-1} \cdot \begin{bmatrix} 1 \\ 1 \end{bmatrix} z \, dz = +\frac{z}{xy} \cdot \begin{bmatrix} 0 & -y \\ -x & x \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 1 \end{bmatrix} dz$$
$$= \frac{z}{xy} \cdot \begin{bmatrix} -y \\ 0 \end{bmatrix} dz = -\frac{z}{x} \cdot \begin{bmatrix} 1 \\ 0 \end{bmatrix} dz$$

Therefore

$$dx = -\frac{z}{x}dz$$
 and $dy = 0 dz$
 $\Rightarrow \frac{\partial x}{\partial z} = -\frac{z}{x}$ and $\frac{\partial y}{\partial z} = 0$

[The intersection is the pair of circles $x^2 + z^2 = 9$, $y = \pm 4$. Because y never changes on each circle, x is actually a function of z only.]

Differentials

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Example 2.3.6

A surface is defined by $f(x, y, z) = xz + y^2 z + z = 1$. Find $\frac{\partial z}{\partial y}$.

Implicit method:

$$df = \frac{\partial f}{\partial x}dx + \frac{\partial f}{\partial y}dy + \frac{\partial f}{\partial z}dz$$

$$0 = z \, dx + 2yz \, dy + (x + y^2 + 1) \, dz$$
But $f = 1 \implies (x + y^2 + 1) \, z = 1$

$$\Rightarrow \frac{dz}{z} = -z \, dx - 2yz \, dy \implies dz = -z^2 (dx + 2y \, dy)$$

In the slice in which the partial derivative $\frac{\partial z}{\partial y}$ is evaluated, x is constant $\Rightarrow dx = 0$.

$$\Rightarrow \frac{\partial z}{\partial y} = -z^2 (0 + 2y) = -2yz^2$$

Explicit method:

$$z = (x + y^{2} + 1)^{-1} \implies \frac{\partial z}{\partial y} = -\underbrace{(x + y^{2} + 1)^{-2}}_{z^{2}}(0 + 2y + 0) = -2yz^{2}$$

In general, if $a \, dx + b \, dy = c \, dz$ then $\frac{\partial z}{\partial x} = \frac{a}{c}$ (because y is constant and dy = 0 in the slice in which $\frac{\partial z}{\partial x}$ is evaluated) and $\frac{\partial z}{\partial y} = \frac{b}{c}$ (because x is constant and dx = 0 in the slice in which $\frac{\partial z}{\partial y}$ is evaluated).

2.4 <u>The Jacobian</u>

The Jacobian is a conversion factor for a differential of area or volume between one orthogonal coordinate system and another.

Let (x, y), (u, v) be related by the pair of simultaneous equations

$$f(x, y, u, v) = c_1$$

$$g(x, y, u, v) = c_2$$

 $\Rightarrow df = f_x dx + f_y dy + f_u du + f_v dv = 0$ and $dg = g_x dx + g_y dy + g_u du + g_v dv = 0$

$$\Rightarrow \underbrace{\begin{pmatrix} f_x & f_y \\ g_x & g_y \end{pmatrix}}_{A} \begin{pmatrix} dx \\ dy \end{pmatrix} = \underbrace{\begin{pmatrix} -f_u & -f_v \\ -g_u & -g_v \end{pmatrix}}_{B} \begin{pmatrix} du \\ dv \end{pmatrix}$$

$$\Rightarrow \begin{pmatrix} dx \\ dy \end{pmatrix} = A^{-1}B \begin{pmatrix} du \\ dv \end{pmatrix}$$

which leads to

$$dA = dx \, dy = \frac{\partial(x, y)}{\partial(u, v)} \, du \, dv$$

where the Jacobian is

$$\frac{\partial(x, y)}{\partial(u, v)} = \left| \frac{\det B}{\det A} \right|$$

The Jacobian for the transformation from (x, y) to (u, v) is also the magnitude of the cross product of the tangent vectors that define the boundaries of the element of area, so that

$$\frac{\partial(x, y)}{\partial(u, v)} = \left| \frac{\partial \vec{\mathbf{r}}}{\partial u} \times \frac{\partial \vec{\mathbf{r}}}{\partial v} \right|$$

Example 2.4.1

Transform the element of area dA = dx dy to plane polar coordinates.

$$x = r \cos \theta, \quad y = r \sin \theta.$$

$$f = x - r \cos \theta = 0$$

$$g = y - r \sin \theta = 0$$

$$\Rightarrow \quad df = dx - \cos \theta \, dr + r \sin \theta \, d\theta = 0$$
and
$$dg = dy - \sin \theta \, dr - r \cos \theta \, d\theta = 0$$

$$\Rightarrow \underbrace{\begin{pmatrix} 1 & 0 \\ 0 & 1 \\ A \end{pmatrix}}_{A} \underbrace{\begin{pmatrix} dx \\ dy \end{pmatrix}}_{B} = \underbrace{\begin{pmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \\ B \\ \end{pmatrix}}_{B} \underbrace{\begin{pmatrix} dr \\ d\theta \\ d\theta \\ \end{pmatrix}}_{B}$$

$$\Rightarrow \frac{\partial(x, y)}{\partial(r, \theta)} = \left| \frac{|B|}{|A|} \right| = \left| \frac{r \cos^2 \theta + r \sin^2 \theta}{1} \right| = |r| \quad (=r \quad \text{if } r \ge 0)$$

Therefore

$$dA = \underline{r \, dr \, d\theta}$$

If x, y can be written as explicit functions of (u, v), then an explicit form of the Jacobian is available:

$$\frac{\partial(x, y)}{\partial(u, v)} = \begin{vmatrix} \frac{\partial x}{\partial u} & \frac{\partial x}{\partial v} \\ \frac{\partial y}{\partial u} & \frac{\partial y}{\partial v} \end{vmatrix}$$

The Jacobian can also be used to express an element of volume in terms of another orthogonal coordinate system:

$$dV = dx \, dy \, dz = \frac{\partial(x, y, z)}{\partial(u, v, w)} \, du \, dv \, dw$$

Spherical Polar Coordinates – a reminder.

The "declination" angle θ is the angle between the positive *z* axis and the radius vector **r**. $0 \le \theta \le \pi$.

The "azimuth" angle ϕ is the angle on the *x-y* plane, measured anticlockwise from the positive *x* axis, of the shadow of the radius vector. $0 \le \phi \le 2\pi$.

$$z = r \cos \theta$$
.

The shadow of the radius vector on the *x*-*y* plane has length $r \sin \theta$.

It then follows that

$$x = r \sin \theta \cos \phi$$
 and $y = r \sin \theta \sin \phi$.

Example 2.4.2

Express the element of volume dV in spherical polar coordinates, $\vec{\mathbf{r}} = \langle x, y, z \rangle = \langle r \sin \theta \cos \phi, r \sin \theta \sin \phi, r \cos \theta \rangle$

Using the explicit form,

$$\frac{\partial(x, y, z)}{\partial(r, \theta, \phi)} = \left\| \begin{vmatrix} x_r & x_\theta & x_\phi \\ y_r & y_\theta & y_\phi \\ z_r & z_\theta & z_\phi \end{vmatrix} \right| = \left\| \begin{aligned} \sin\theta\cos\phi & r\cos\theta\cos\phi & -r\sin\theta\sin\phi \\ \sin\theta\sin\phi & r\cos\theta\sin\phi & r\sin\theta\cos\phi \\ \cos\theta & -r\sin\theta & 0 \end{aligned} \right|$$

Expanding along the last row of the determinant,

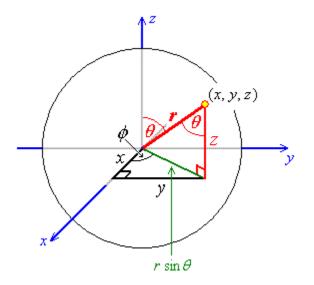
$$\frac{\partial(x, y, z)}{\partial(r, \theta, \phi)} = \left| \cos \theta \left(r^2 \sin \theta \cos \theta \right) \left(\cos^2 \phi + \sin^2 \phi \right) + r \sin \theta \left(r \sin^2 \theta \right) \left(\cos^2 \phi + \sin^2 \phi \right) + 0 \right|$$

$$= |r^{2}\sin\theta (\cos^{2}\theta + \sin^{2}\theta)| = r^{2}\sin\theta$$

Note that $\sin \theta \ge 0$ because $0 \le \theta \le \pi$.

Therefore

$$dV = r^2 \sin \theta \, dr \, d\theta \, d\phi$$



The Jacobian

For a transformation from Cartesian to plane polar coordinates in \mathbb{R}^2 ,

$$\frac{\partial(x, y)}{\partial(r, \theta)} = r \text{ so that } dA = dx \, dy = r \, dr \, d\theta$$

For cylindrical polar coordinates in \mathbb{R}^3 ,
$$\frac{\partial(x, y, z)}{\partial(r, \theta, z)} = r \implies dV = r \, dr \, d\theta \, dz$$

Explicit method for plane polar coordinates:

$$x = r \cos \theta, \quad y = r \sin \theta$$

$$\Rightarrow \frac{\partial(x, y)}{\partial(r, \theta)} = ABS \left[\det \begin{bmatrix} \frac{\partial x}{\partial r} & \frac{\partial x}{\partial \theta} \\ \frac{\partial y}{\partial r} & \frac{\partial y}{\partial \theta} \end{bmatrix} \right] = \left\| \cos \theta - r \sin \theta \\ \sin \theta & r \cos \theta \\ \right\|$$

$$= r \cos^2 \theta + r \sin^2 \theta = r.$$

Implicit method for plane polar coordinates:

Implicit method for plane polar coordinates: [which is a repeat of page 2-17.]

$$f = x - r \cos \theta = 0 \implies df = dx - \cos \theta \, dr + r \sin \theta \, d\theta = 0$$

$$g = y - r \sin \theta = 0 \implies dg = dy - \sin \theta \, dr - r \cos \theta \, d\theta = 0$$

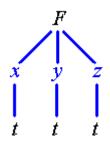
$$\implies \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} dx \\ dy \end{bmatrix} = \begin{bmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \end{bmatrix} \cdot \begin{bmatrix} dr \\ d\theta \end{bmatrix}$$

$$\implies \frac{\partial(x, y)}{\partial(r, \theta)} = \left| \frac{|\mathbf{B}|}{|\mathbf{A}|} \right| = \left| \frac{r \cos^2 \theta + r \sin^2 \theta}{1} \right| = |r| = r$$

2.5 <u>The Gradient Vector</u>

Let
$$F = F(\mathbf{r})$$
 and $\mathbf{r} = \mathbf{r}(t)$.
$$\frac{dF}{dt} = \frac{\partial F}{\partial x}\frac{dx}{dt} + \frac{\partial F}{\partial y}\frac{dy}{dt} + \frac{\partial F}{\partial z}\frac{dz}{dt}$$

$$=\left(\left\langle\frac{\partial}{\partial x},\frac{\partial}{\partial y},\frac{\partial}{\partial z}\right\rangle F\right)\cdot\frac{d\vec{\mathbf{r}}}{dt}$$



$$\frac{dF}{dt} = \bar{\nabla}F \cdot \frac{d\bar{\mathbf{r}}}{dt}$$

 $\vec{\nabla}$ = "del" or "nabla" = gradient operator

 $\overline{\nabla}F$ = gradient vector

Directional Derivative

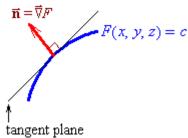
The rate of change of the function F at the point P_0 in the direction of the vector $\vec{a} = a\hat{a} (a \neq 0)$ is

$$D_{\hat{\mathbf{a}}}F|_{P_{o}} = \overline{\nabla}F|_{P_{o}}\cdot\hat{\mathbf{a}}$$

The maximum value of the directional derivative of F at any point P_0 occurs when the vector **a** is parallel to the gradient vector $\vec{\nabla}F$.

Also, the vector $\mathbf{\bar{N}} = \mathbf{\nabla}F$ is normal to the surface defined by $F(\mathbf{r}) = \text{constant}$.

[Interpretation of the gradient vector: The gradient vector is in the plane of the dependent variables.



Its direction at any point is the direction in which one must travel in order to experience the greatest possible rate of increase of the dependent variable at that point. Its magnitude is that greatest possible rate of increase.]

Example 2.5.1

The temperature in a region within 10 units of the origin follows the form $T = r e^{-r}$, where $r = \sqrt{x^2 + y^2 + z^2}$. Find the rate of temperature change at the point (-1, -1, +1) in the direction of the vector $\langle 1, 0, 0 \rangle$.

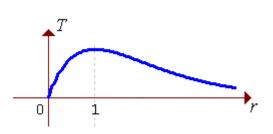
T and r are symmetric with respect to (x, y, z).

$$\frac{\partial T}{\partial x} = \frac{dT}{dr} \frac{\partial r}{\partial x}$$

$$\frac{dT}{dr} = \frac{d}{dr} (r e^{-r}) = (1-r)e^{-r}$$

$$\frac{\partial r}{\partial x} = \frac{1}{2} (x^2 + y^2 + z^2)^{-1/2} (2x + 0 + 0) = \frac{x}{r}$$

$$\Rightarrow \frac{\partial T}{\partial x} = (1-r)e^{-r} \frac{x}{r}$$



By symmetry, it then follows that

$$\frac{\partial T}{\partial y} = (1-r)e^{-r}\frac{y}{r}$$
 and $\frac{\partial T}{\partial z} = (1-r)e^{-r}\frac{z}{r}$

$$\Rightarrow \quad \overline{\nabla}T = \frac{1-r}{r} e^{-r} \langle x, y, z \rangle$$

$$\hat{\mathbf{a}} = \hat{\mathbf{i}} \implies D_{\hat{\mathbf{a}}}T = \overline{\nabla}T\cdot\hat{\mathbf{i}} = \frac{1-r}{r}e^{-r}\langle x, y, z\rangle\cdot\langle 1, 0, 0\rangle$$

At
$$(-1, -1, +1)$$
, $r = \sqrt{1+1+1} = \sqrt{3}$

$$\therefore \quad D_{\hat{a}}T|_{P} = \frac{1-\sqrt{3}}{\sqrt{3}}e^{-\sqrt{3}}\left\langle -1, -1, +1\right\rangle \cdot \left\langle 1, 0, 0\right\rangle = \frac{\sqrt{3}-1}{\sqrt{3}}e^{-\sqrt{3}} \approx 0.0748 \,\mathrm{K\,m^{-1}}$$

[Note that, in this example, the temperature distribution is spherically symmetric about the origin, with a minimum of zero at the origin, rising to a maximum of 1/e = 0.368... one metre away from the origin and falling asymptotically back to zero as $r \rightarrow \infty$. Outside the r = 1 sphere, the gradient vector points radially inwards. At (-1, -1, 1) the gradient vector has a magnitude of approximately 0.104 K m⁻¹.]

Central Force Law

If $\phi(\mathbf{r})$ is the **potential function** for some force per unit mass or force per unit charge $\mathbf{F}(\mathbf{r})$, then $\mathbf{F} = -\nabla \phi$. For a central force law, the potential function is spherically symmetric and is dependent only on the distance r from the origin. When the central force law is a simple power law, $\phi = k r^n$.

$$r = \sqrt{x^{2} + y^{2} + z^{2}}$$

$$\Rightarrow \frac{\partial \phi}{\partial x} = \frac{d\phi}{dr} \frac{\partial r}{\partial x}$$

$$= knr^{n-1} \cdot \frac{1}{2} (x^{2} + y^{2} + z^{2})^{-1/2} (2x + 0 + 0)$$

$$= knx \cdot \frac{r^{n-1}}{r} = knxr^{n-2}$$

By symmetry, $\frac{\partial \phi}{\partial y} = kn y r^{n-2}$ and $\frac{\partial \phi}{\partial z} = kn z r^{n-2}$

$$\Rightarrow \quad \vec{\nabla}\phi = knr^{n-2}\langle x, y, z \rangle = knr^{n-2}\vec{\mathbf{r}}$$

But $\mathbf{\bar{r}} = r \, \hat{\mathbf{r}}$

Therefore

$$-\vec{\nabla}\phi = \vec{\mathbf{F}} = -knr^{n-1}\hat{\mathbf{r}}$$

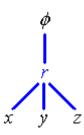
Examples include the inverse square laws, for which n = -1:

Electromagnetism:

$$k = \frac{Q}{4\pi\varepsilon}, \quad \phi = \frac{Q}{4\pi\varepsilon r}, \quad \mathbf{\bar{F}} = -\mathbf{\bar{\nabla}}\phi = \frac{Q}{4\pi\varepsilon r^2}\hat{\mathbf{r}}$$

Gravity:

$$k = -GM$$
, $\phi = \frac{-GM}{r}$, $\vec{\mathbf{F}} = -\vec{\nabla}\phi = \frac{-GM}{r^2}\hat{\mathbf{r}}$



Applications of the Directional Derivative and the Gradient:

- (1) The directional derivative D at the point $P_0 = (x_0, y_0, z_0)$ is maximized by choosing **a** to be parallel to ∇F at P_0 , so that $D_{\hat{\mathbf{a}}}F|_{P_0} = |\nabla F(\vec{\mathbf{r}})|_{P_0}$.
- (2) A normal vector to the surface F(x, y, z) = c at the point $P_0 = (x_0, y_0, z_0)$ is $\vec{\mathbf{N}} = [\vec{\nabla}F(\vec{\mathbf{r}})]_{P_0}$.
- (3) The equation of the line normal to the surface F(x, y, z) = c at the point $P_o = (x_o, y_o, z_o)$ is $\vec{\mathbf{r}} = \left[\overrightarrow{OP_o}\right] + t \left[\overrightarrow{\nabla}F\right]_{P_o} = \vec{\mathbf{r}}_o + t \left[\overrightarrow{\nabla}F\right]_{P_o}$
- (4) The equation of the tangent plane to the surface F(x, y, z) = c at the point $P_0 = (x_0, y_0, z_0)$ is $\left[\vec{\mathbf{r}} \overrightarrow{OP_o}\right] \bullet \left[\vec{\nabla}F\right]_{P_o} = \left[\vec{\mathbf{r}} \vec{\mathbf{r}}_o\right] \bullet \left[\vec{\nabla}F\right]_{P_o} = 0$
- (5) If the point $P_0 = (x_0, y_0, z_0)$ lies on both of the surfaces F(x, y, z) = c and G(x, y, z) = k, then the angle of intersection θ of the surfaces at the point is given by

$$\cos \theta = \frac{\left[\bar{\nabla}F\right]_{P_o} \bullet \left[\bar{\nabla}G\right]_{P_o}}{\left|\bar{\nabla}F\right|_{P_o} \left|\bar{\nabla}G\right|_{P_o}}$$

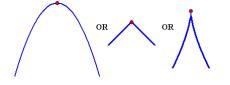
2.6 <u>Maxima and Minima</u>

Much of differential calculus in the study of maximum and minimum values of a function of one variable carries over to the case of a function of two (or more) variables. In order to visualize what is happening, we shall restrict our attention to the case of functions of two variables, z = f(x, y).

For a function f(x, y) defined on some domain D in \mathbb{R}^2 , the point $P(x_0, y_0)$ is a **critical point** [and the value $f(x_0, y_0)$ is a **critical value**] of f if

- 1) P is on any boundary of D; or
- 2) $f(x_0, y_0)$ is undefined; or
- 3) f_x and/or f_y is undefined at P; or
- 4) f_x and f_y are both zero at $P \implies \nabla f = \mathbf{0}$ at P).

A local maximum continues to be equivalent to a "hilltop",



while a local minimum continues to be equivalent to a "valley bottom".

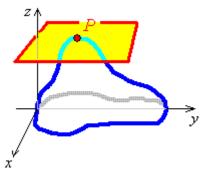


"Local extremum" is a collective term for local maximum or local minimum.

Instead of tangent lines being horizontal at critical points of type (4), we now have tangent *planes* being horizontal at critical points of type (4).

At any local extremum at which f(x, y) is differentiable, the tangent plane is horizontal, $f_x = f_y = 0$ and $\nabla f = \mathbf{0}$. The converse is false.

 $\nabla f = 0$ does not guarantee a local extremum. There could be a saddle point (the higher dimensional equivalent of a point of inflection) instead.



Example 2.6.1

Find and identify the nature of the extrema of $f(x, y) = x^2 + y^2 + 4x - 6y$.

Polynomial functions of x and y are defined and are infinitely differentiable on <u>all</u> of \mathbb{R}^2 . Therefore the only critical points are of type (4).

$$\vec{\nabla}f = \langle 2x + 4, 2y - 6 \rangle$$

 $\vec{\nabla}f = \vec{0} \implies (x, y) = (-2, 3)$ only.

But
$$f(x, y) = (x+2)^2 - 4 + (y-3)^2 - 9$$

 $\ge 0 - 4 + 0 - 9 = -13 = f(-2, 3)$

Therefore the only local extremum is a minimum value of -13 at (-2, 3).

It is also an **absolute minimum**.

To determine the nature of a critical point:

- 1) Examine the values of f in the neighbourhood of P; or
- 2) [First derivative test:] Examine the changes in f_x and f_y at P; or
- 3) Use the <u>second derivative test</u>:

At all points (a, b) where $\nabla f = \mathbf{0}$, find all second partial derivatives, then find

$$D = \begin{vmatrix} f_{xx} & f_{xy} \\ f_{yx} & f_{yy} \end{vmatrix}$$

and evaluate D at (x, y) = (a, b).

D(a, b) > 0 and $f_{xx}(a, b) > 0 \implies$ a relative **minimum** of f is at (a, b)D(a, b) > 0 and $f_{xx}(a, b) < 0 \implies$ a relative **maximum** of f is at (a, b) $D(a, b) < 0 \implies$ a **saddle point** of f is at (a, b) $D(a, b) = 0 \implies$ test fails (no information). Example 2.6.1 (again):

Find all extrema of $f(x, y) = x^2 + y^2 + 4x - 6y$.

Any critical points will be of type (4) only.

$$\vec{\nabla} f = \langle 2x+4, 2y-6 \rangle = \vec{0} \implies (x, y) = (-2, 3)$$
 only.

$$f_{xx} = 2, f_{xy} = 0, f_{yy} = 2$$

$$\Rightarrow D = \begin{vmatrix} 2 & 0 \\ 0 & 2 \end{vmatrix} = 4 > 0 \quad \forall (x, y)$$

D > 0 and $f_{xx} > 0$ at $(-2, 3) \Rightarrow$ there is a relative minimum at (-2, 3) and the minimum value is f(-2, 3) = -13.

As there are no other critical points, f(x, y) has an **absolute minimum** value of -13 at (-2, 3) and has no maxima.

[z = f(x, y) is a circular paraboloid, vertex at (-2, 3, -13) and axis of symmetry parallel to the *z*-axis.]

<u>Example 2.6.2</u> Find all extrema of $f(x, y) = 2x^3 + xy^2 + 5x^2 + y^2$.

Any critical points will be of type (4) only.

$$\overline{\nabla}f = \langle 6x^2 + y^2 + 10x, \ 2xy + 2y \rangle$$

$$\overline{\nabla}f = \overline{\mathbf{0}} \implies 6x^2 + 10x + y^2 = 0 \quad (1)$$

$$2y (x + 1) = 0 \quad (2)$$
(2) $\implies y = 0 \text{ or } x = -1$

$$y = 0 \text{ in } (1) \implies 2x (3x + 5) = 0$$

$$\implies x = 0 \text{ or } x = -\frac{5}{3}$$

$$x = -1 \text{ in } (1) \implies 6 - 10 + y^2 = 0$$

$$\implies y^2 = 4 \implies y = \pm 2$$
Critical points are $(x, y) = (0, 0), (-5/3, 0), (-1, -2) \text{ and } (-1, +2).$

$$f_x = 6x^2 + 10x + y^2 \quad \text{and} \quad f_y = 2y (x + 1)$$

$$\implies f_{xx} = 12x + 10$$

$$f_{xy} = 2y$$
and
$$f_{yy} = 2 (x + 1)$$

Example 2.6.2 (continued)

$$(0,0): z = f(0,0) = 0$$

$$D = \begin{vmatrix} 10 & 0 \\ 0 & 2 \end{vmatrix} = 20 > 0$$

D > 0 and $f_{xx} = 10 > 0 \implies$ minimum

$$(-\frac{5}{3},0): \qquad z = 2\left(-\frac{5}{3}\right)^3 + 0 + 5\left(-\frac{5}{3}\right)^2 + 0$$

$$= \left(-\frac{5}{3}\right)^2 \left(2\left(-\frac{5}{3}\right) + 5\right) = \frac{25}{9}\left(\frac{-10+15}{3}\right) = \frac{125}{27}$$

$$D = \begin{vmatrix} 12\left(-\frac{5}{3}\right) + 10 & 0 \\ 0 & 2\left(-\frac{5}{3} + 1\right) \end{vmatrix} = \begin{vmatrix} -10 & 0 \\ 0 & -\frac{4}{3} \end{vmatrix} = +\frac{40}{3} > 0$$

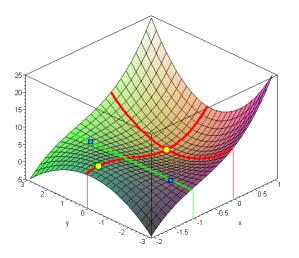
D > 0 and $f_{xx} < 0 \implies$ maximum

$$(-1, \pm 2): \qquad z = f(-1, \pm 2) = 2(-1) + (-1)(\pm 2)^2 + 5(-1)^2 + (\pm 2)^2$$
$$= -2 - 4 + 5 + 4 = 3$$
$$D = \begin{vmatrix} -2 & \pm 4 \\ \pm 4 & 0 \end{vmatrix} = 0 - (\pm 4)^2 = -16$$

$D < 0 \implies$ saddle point

Therefore the critical points are

Minimum at (0, 0, 0)Maximum at $\left(-\frac{5}{3}, 0, \frac{125}{27}\right)$ Saddle points at (-1, -2, 3) and (-1, 2, 3)

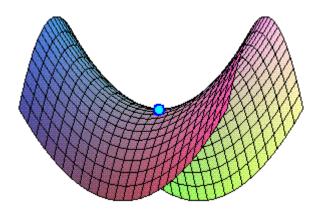


Example 2.6.3 Find all extrema of $z = f(x, y) = x^2 - y^2$ $\vec{\nabla}f = \langle 2x, -2y \rangle$ $\vec{\nabla}f = \vec{0} \implies$ (x, y) = (0, 0) only. f(0, 0) = 0

$$D = \begin{vmatrix} f_{xx} & f_{xy} \\ f_{yx} & f_{yy} \end{vmatrix} = \begin{vmatrix} 2 & 0 \\ 0 & -2 \end{vmatrix} < 0$$

Therefore the only critical point is a <u>saddle point at (0, 0, 0)</u>.

The surface is an hyperbolic paraboloid.



Example 2.6.4

Find and identify the nature of the extrema of $f(x, y) = \sqrt{1 - x^2 - y^2}$.

The domain of f(x, y) is the circle $x^2 + y^2 = 1$ and all points inside it.

 $f(x,y) = (1 - x^2 - y^2)^{1/2}$

$$\Rightarrow \quad \bar{\nabla}f(x,y) = \frac{1}{2} \left\langle \frac{-2x}{\sqrt{1-x^2-y^2}}, \frac{-2y}{\sqrt{1-x^2-y^2}} \right\rangle = -\frac{1}{\sqrt{1-x^2-y^2}} \langle x, y \rangle$$

Inside the domain, f and its partial derivatives are well defined everywhere, except that the gradient diverges everywhere on the boundary. Therefore the only critical points to consider are types (1), (3) and (4).

Type (4) [zero gradient]: Inside the domain, $\nabla f = \mathbf{0} \implies (x, y) = (0, 0)$ only. f(0, 0) = 1

However, $0 \le 1 - x^2 - y^2 < 1$ everywhere else in the domain $\Rightarrow f(x, y) < f(0, 0)$ everywhere else in the domain. \Rightarrow there is an absolute maximum at (0, 0, 1).

Types (3) [undefined gradient] and (1) [the boundary]: Everywhere on the boundary, $1 - x^2 - y^2 = 1 - 1 = 0$ and 0 < f(x, y) everywhere else in the domain.

 \Rightarrow the <u>absolute minima are all points on the circle $x^2 + y^2 = 1$, z = 0</u>.

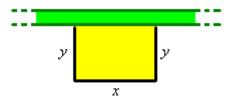
[The surface is the upper half of the sphere $x^2 + y^2 + z^2 = 1$.]

2.7 Lagrange Multipliers

In order to obtain an appreciation for the geometric foundation for the method of Lagrange multipliers, we shall begin with an example that could be solved in another way.

Example 2.7.1

A farmer wishes to enclose a rectangle of land. One side is a straight hedge, more than 30 m long. The farmer has a total length of 12 m of fencing available for the other three sides. What is the greatest area that can be enclosed by the available fencing?



The function to be maximized is the area enclosed by the fencing and the hedge:

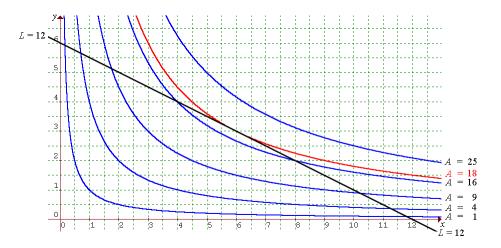
A(x, y) = x y

The constraint is the length of fencing available:

L(x, y) = x + 2y = 12

There are additional constraints. Neither length may be a negative number. Therefore any solution is confined to the first quadrant of the (x, y) graph.

Superimpose the graph of the constraint function L(x, y) = 12 on the contour graph of the function z = A(x, y):



The least value of A(x, y) is zero, on the coordinate axes. At (0, 6), two 6-metre segments of fence are touching, enclosing zero area. At (12, 0), all of the fence is up against the hedge, enclosing zero area.

As one travels along the constraint line L = 12 from one absolute minimum at (0, 6) to the other absolute minimum at (12, 0), the line first passes through increasing contours of A(x, y),

$$A(6 - \sqrt{34}, 3 + \sqrt{8.5}) = 1$$
$$A(6 - 2\sqrt{7}, 3 + \sqrt{7}) = 4$$
$$A(6 - \sqrt{18}, 3 + \sqrt{4.5}) = 9$$
$$A(4, 4) = 16$$

until, somewhere in the vicinity of (6, 3), the area reaches a maximum value, then declines, (for example, A(8, 2) = 16), back to the other absolute minimum at (12, 0).

At the maximum, the graph of L(x, y) = 12 just touches a contour of A(x, y).

The two graphs share a common tangent there.

$$\Rightarrow$$
 gradients are parallel

$$\Rightarrow \quad \overline{\nabla}A = \lambda \,\overline{\nabla}L$$

 $\Rightarrow \langle y, x \rangle = \lambda \langle 1, 2 \rangle$

Therefore solve the system of simultaneous equations

$$y = \lambda$$

 $x = 2\lambda$
 $x + 2y = 12$ (the constraint)

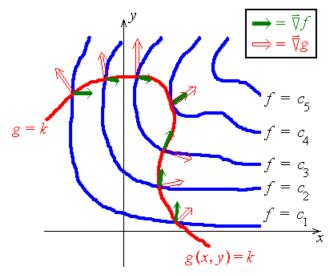
Solution: (x, y) = (6, 3).

From the graph, A is a maximum at (6, 3).

$$A_{\text{max}} = A(6, 3) = 6 \times 3 = \underline{18 \text{ m}^2}.$$

 λ is the Lagrange multiplier.

To maximize f(x, y) subject to the constraint g(x, y) = k:



As one travels along the constraint curve g(x, y) = k, the maximum value of f(x, y) occurs only where the two gradient vectors are parallel. This principle can be extended to the case of functions of more than two variables.

General Method of Lagrange Multipliers

To find the maximum or minimum value(s) of a function $f(x_1, x_2, ..., x_n)$ subject to a constraint $g(x_1, x_2, ..., x_n) = k$, solve the system of simultaneous (usually non-linear) equations in (n + 1) unknowns:

$$\nabla f = \lambda \nabla g$$
$$g = k$$

where λ is the Lagrange multiplier.

Then identify which solution(s) gives a maximum or minimum value for f.

This can also be extended to the case of more than one constraint:

In the presence of two constraints $g(x_1, x_2, ..., x_n) = k$ and $h(x_1, x_2, ..., x_n) = c$, solve the system in (n + 2) unknowns:

$$\nabla f = \lambda \nabla g + \mu \nabla h$$
$$g = k$$
$$h = c$$

Example 2.7.2

Find the points, on the curve of intersection of the surfaces x + y - 2z = 6 and $2x^2 + 2y^2 = z^2$, that are nearest / farthest from the origin.

The maximum and minimum values of distance d occur at the same place as the maximum and minimum values of d^2 . [Differentiation of d^2 will be much easier than differentiation of d.]

The function to be maximized/minimized is

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

The constraints are

and

 $h(x, y, z) = 2x^2 + 2y^2 - z^2 = 0$

g(x, y, z) = x + y - 2z = 6

$$\overline{\nabla}f = \lambda \,\overline{\nabla}g + \mu \,\overline{\nabla}h$$

$$\Rightarrow \langle 2x, 2y, 2z \rangle = \lambda \langle 1, 1, -2 \rangle + \mu \langle 4x, 4y, -2z \rangle$$

$$\Rightarrow 2x(1 - 2\mu) = \lambda \qquad (1)$$

$$2y(1 - 2\mu) = \lambda \qquad (2)$$

$$2z(1 + \mu) = -2\lambda \qquad (3)$$

$$x + y - 2z = 6 \qquad (4)$$
and
$$2x^2 + 2y^2 - z^2 = 0 \qquad (5)$$

$$(1) - (2) \Rightarrow 2(x - y)(1 - 2\mu) = 0$$

$$\Rightarrow y = x \text{ or } 2\mu = 1.$$

$$Case \ 2\mu = 1:$$

$$(1) \text{ or } (2) \Rightarrow \lambda = 0$$

$$(3) \Rightarrow z = 0 \quad [\text{or } \mu = -1, \text{ which contradicts } 2\mu = (5) \Rightarrow 2(x^2 + y^2) = 0 \Rightarrow (x, y) = (0, 0) \text{ only.}$$

Note that *f*, *g* and *h* are all symmetric with respect to (x, y). It then follows that, at any critical point, y = x.

Equation (2) is redundant, which allows for a faster solution than is shown here.

= 1.]

Example 2.7.2 (continued)

But (4) $\Rightarrow 0+0-0 = 6$ - impossible!

Therefore $2\mu \neq 1$.

<u>Case y = x</u>:

 $(4) \Rightarrow 2x - 2z = 6 \Rightarrow z = x - 3$ $(5) \Rightarrow 4x^{2} - (x - 3)^{2} = 0$ $\Rightarrow 4x^{2} - x^{2} + 6x - 9 = 0$ $\Rightarrow 3(x^{2} + 2x - 3) = 0$ $\Rightarrow 3(x + 3) (x - 1) = 0$ $\Rightarrow x = -3 \text{ or } x = 1.$ $x = -3 \Rightarrow y = -3$ $(6) \Rightarrow z = -6$ $f(-3, -3, -6) = (-3)^{2} + (-3)^{2} + (-6)^{2}$ $\Rightarrow d^{2} = 9(1 + 1 + 4) = 54.$ $x = 1 \Rightarrow y = 1$ $(6) \Rightarrow z = -2$ $f(1, 1, -2) = 1^{2} + 1^{2} + (-2)^{2}$ $\Rightarrow d^{2} = 1 + 1 + 4 = 6.$

These are the only critical points. Therefore <u>The nearest point is (1, 1, -2) and the farthest point is (-3, -3, -6).</u>

[The surfaces are a plane and a right circular cone. The curve of intersection is an ellipse.]

2.8 <u>Miscellaneous Additional Examples</u>

Example 2.8.1

A window has the shape of a rectangle surmounted by a semicircle. Find the maximum area of the window, when the perimeter is constrained to be 8 m.

The function to be maximized is

$$A(r,h) = 2rh + \frac{1}{2}\pi r^2$$

The constraint is the perimeter function

$$P(r,h) = h + 2r + h + \pi r = 2h + (\pi + 2)r = 8$$

Single variable method:

$$P = 8 \qquad \Rightarrow \quad h = \frac{8 - (\pi + 2)r}{2}$$
$$\Rightarrow \quad A(r) = r\left(8 - (\pi + 2)r\right) + \frac{1}{2}\pi r^{2} = 8r - \left(\frac{\pi + 4}{2}\right)r^{2}$$

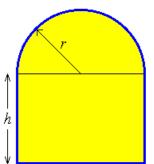
$$\Rightarrow \frac{dA}{dr} = 8 - (\pi + 4)r$$

$$\frac{dA}{dr} = 0 \implies r = \frac{8}{\pi + 4} \text{ only.}$$

The only other critical points are on the domain boundaries, at r = 0 (when A = 0, clearly an absolute minimum) and at h = 0 (when $A = \frac{1}{2}\pi r^2$).

Inside the domain, r > 0 and $h > 0 \implies A > 0$ (and A is differentiable throughout).

Therefore A must be at either a maximum or a point of inflexion at $r = \frac{8}{\pi + 4}$.



Example 2.8.1 (continued)

$$\frac{d^2A}{dr^2} = 0 - (\pi + 4) < 0 \implies \text{maximum at } r = \frac{8}{\pi + 4}.$$

At the maximum,
$$h = \frac{8 - (\pi + 2)r}{2} = \frac{8}{2} \left(\frac{(\pi + 4) - (\pi + 2)}{\pi + 4} \right) = \frac{8}{\pi + 4}$$

$$\Rightarrow A_{\text{max}} = \frac{8 \times 8}{\pi + 4} - \left(\frac{\pi + 4}{2} \right) \left(\frac{8}{\pi + 4} \right)^2 = \frac{32}{\underline{\pi + 4}} \approx 4.48 \text{ m}^2$$

The maximum area occurs when

$$r = h = \frac{8}{\frac{\pi + 4}{2}} \approx 1.12 \,\mathrm{m} \;.$$

Lagrange Multiplier Method:

$$A(r,h) = 2rh + \frac{1}{2}\pi r^{2}$$

$$P(r,h) = h + 2r + h + \pi r = 2h + (\pi + 2)r = 8$$

$$\bar{\nabla}A = \lambda \,\bar{\nabla}P$$

$$\Rightarrow \langle 2h + \pi r, \, 2r \rangle = \lambda \langle \pi + 2, \, 2 \rangle$$

$$\Rightarrow \qquad 2h + \pi r = \lambda (\pi + 2) \qquad (1)$$

$$2r = 2\lambda \qquad (2)$$
and
$$2h + (\pi + 2)r = 8 \qquad (3)$$

and $2h + (\pi + 2)r = 8$

Example 2.8.1 (continued)

(2) $\Rightarrow \lambda = r \Rightarrow$ (1) becomes

$$2h + \pi r = r(\pi + 2)$$

$$\Rightarrow 2h = 2r$$

$$\Rightarrow h = r.$$

(3) \Rightarrow $(\pi+4)r = 8$

Therefore $r = h = \frac{8}{\underline{\pi + 4}} \approx 1.12 \text{ m}$.

At this critical point, $A = \frac{8 \times 8}{\pi + 4} - \left(\frac{\pi + 4}{2}\right) \left(\frac{8}{\pi + 4}\right)^2 = \frac{32}{\pi + 4} \approx 4.48 \text{ m}^2$

The ends of the constraint line are at $(r,h) = B(0,4), C\left(\frac{8}{\pi+2},0\right)$.

At
$$B$$
, $A = A_{\min} = 0$

At C,
$$A = \frac{1}{2}\pi \left(\frac{8}{\pi+2}\right)^2 = \frac{32\pi}{(\pi+2)^2} \approx 3.80 \text{ m}^2$$

A > 0 and differentiable at all points between B and C.

Therefore the maximum area of $\frac{32}{\pi + 4}$ occurs at $r = h = \frac{8}{\frac{\pi + 4}{2}} \approx 1.12 \text{ m}$.

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Example 2.8.2

Find the local and absolute extrema of the function

$$f(x,y) = \sqrt[3]{x^2 + y^2}.$$

$$\frac{\partial f}{\partial x} = \frac{1}{3} \left(x^2 + y^2 \right)^{-2/3} \left(2x + 0 \right) = \frac{2x}{3\sqrt[3]{\left(x^2 + y^2 \right)^2}}$$

By symmetry,

$$\frac{\partial f}{\partial y} = \frac{2y}{3\sqrt[3]{\left(x^2 + y^2\right)^2}}$$

f is continuous on all of \mathbb{R}^2 ,

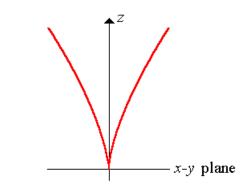
(therefore there is no boundary to check and no critical points of types (1) or (2)), but $|f_x| \to \infty$ and $|f_y| \to \infty$ as $(x, y) \to (0, 0)$ (critical point of type (3)). $f_x \neq 0$ and $f_y \neq 0$ everywhere else (no critical points of type (4)).

Therefore (0, 0) is the only critical point.

But f(0, 0) = 0 and f(x, y) > 0 for all (x, y) except (0, 0).

Therefore the only critical point is an <u>absolute minimum value of 0 at (0, 0)</u>.

[There is a cusp at the origin.]



[Any vertical cross-section containing the *z* axis.]

Example 2.8.3

Find the maximum and minimum values of the function $V(x, y) = 48xy - 32x^3 - 24y^2$ in the unit square $0 \le x \le 1$, $0 \le y \le 1$.

[The full solution to this question is available on the course web site, at "www.engr.mun.ca/~ggeorge/2422/notes/c2maxex1.html".]

 $-32 \leq V(x, y) \leq +2,$

with the absolute maximum of V at $(x, y) = (\frac{1}{2}, \frac{1}{2})$

and the absolute minimum of V at (x, y) = (1, 0).

Example 2.8.4

A hilltop is modelled by the part of the elliptic paraboloid

$$h(x,y) = 4000 - \frac{x^2}{1000} - \frac{y^2}{250}$$

that is above the x-y plane. At the point P(500, 300, 3390), in which direction is the steepest ascent?

$$\bar{\nabla}h = \left\langle \frac{-x}{500}, \frac{-y}{125} \right\rangle$$

 $\Rightarrow \quad \bar{\nabla}h\Big|_{P} = \left\langle -1, -\frac{12}{5} \right\rangle$

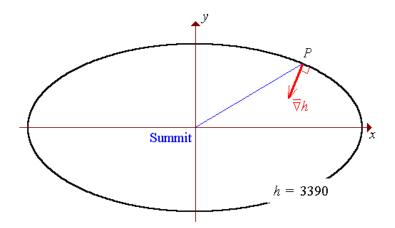
Therefore the steepest ascent is in the direction of the vector

$$\underline{\langle -5, -12 \rangle}$$

Note that this vector does not point directly at the summit.

The summit is on the *z* axis.

From the point *P*(500, 300, 3390), the summit is in the direction $\langle -5, -3 \rangle$.



Example 2.8.5

Show that
$$u(x,t) = \frac{1}{2} (f(x-ct) + f(x+ct))$$
 satisfies $\frac{\partial^2 u}{\partial t^2} = c^2 \frac{\partial^2 u}{\partial x^2}$.

Let r = x - ct

and s = x + ct

Then

$$u = \frac{1}{2} \left(f(r) + f(s) \right)$$

$$\bigwedge_{x = t}^{r} \bigwedge_{x = t}^{S}$$

$$\frac{\partial u}{\partial t} = \frac{\partial u}{\partial r} \cdot \frac{\partial r}{\partial t} + \frac{\partial u}{\partial s} \cdot \frac{\partial s}{\partial t}$$
$$= \frac{1}{2} \frac{df}{dr} (-c) + \frac{1}{2} \frac{df}{ds} (+c)$$
$$\Rightarrow u_{tt} = \frac{\partial u_{t}}{\partial r} \cdot \frac{\partial r}{\partial t} + \frac{\partial u_{t}}{\partial s} \cdot \frac{\partial s}{\partial t}$$

$$= \frac{1}{2} \frac{d^2 f}{dr^2} (-c)^2 + \frac{1}{2} \frac{d^2 f}{ds^2} (+c)^2 = \frac{c^2}{2} \left(\frac{d^2 f}{dr^2} + \frac{d^2 f}{ds^2} \right)$$

$$\frac{\partial u}{\partial x} = \frac{\partial u}{\partial r} \cdot \frac{\partial r}{\partial x} + \frac{\partial u}{\partial s} \cdot \frac{\partial s}{\partial x} = \frac{1}{2} \frac{df}{dr} \times 1 + \frac{1}{2} \frac{df}{ds} \times 1$$
$$\implies u_{xx} = \frac{1}{2} \frac{d^2 f}{dr^2} (1)^2 + \frac{1}{2} \frac{d^2 f}{ds^2} (1)^2 = \frac{1}{c^2} u_{tt}$$

Therefore $u_{tt} = c^2 u_{xx}$.

[Thus $u(x, t) = \frac{1}{2}(f(x - ct) + f(x + ct))$ satisfies the wave equation - and f(r) can be *any* twice differentiable function of r !

This function models a pair of identical wave forms, moving at speed c in opposite directions.]

[End of Chapter 2]