

# Introduction

Differentiation has a **geometric** aspect (measuring the *gradient* of an arc and then letting  $\Delta t$  tend to zero) and an **algebraic** aspect (linear approximation:  $f(x) = f(a) + \frac{df}{dx}(a)(x-a) + a \text{ bit}$ ).  
**Recalling**,  $\mathbf{R}^m$  is an *m-dimensional* Euclidean space:  $\underline{x} = (x_1, \dots, x_m)$ ,  $\underline{x} \cdot \underline{y} = x_1y_1 + x_2y_2 + \dots + x_my_m = \sum_{i=1}^m x_iy_i$ . The **norm** is  $|\underline{x}| = (\underline{x} \cdot \underline{x})^{1/2} = (\sum_{i=1}^m x_i^2)^{1/2}$ . The *distance*  $|\underline{x} - \underline{y}|$  is the distance between  $x$  and  $y$ .  $T: \mathbf{R}^m \rightarrow \mathbf{R}^n$  is a *linear* function, with  $T(r\underline{x} + s\underline{y}) = rT(\underline{x}) + sT(\underline{y})$ .

If we are given **bases** for  $\mathbf{R}^m$  and  $\mathbf{R}^n$ , we can work out  $T$  in terms of a *matrix*. A plane through  $O$  in  $\mathbf{R}^3$  has equation  $\underline{a} \cdot \underline{x} = 0$ .  $\underline{a} = (a_1, a_2, a_3)$  is **normal** to the plane. For a plane going through a *vector*  $\underline{b}$ , the equation is  $\underline{a} \cdot (\underline{x} - \underline{b}) = 0$ . Again,  $\underline{a}$  is in the **normal** direction. This generalises to  $\mathbf{R}^n$  for *all*  $n$ .

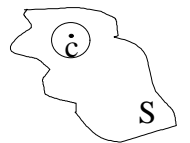
## Section 1: Multivariable Differential Calculus

Let  $S$  be a subset of  $\mathbf{R}^n$ ; and let  $f$  be a *function* from  $S$  to  $\mathbf{R}^n$ ,  $f: S \rightarrow \mathbf{R}^n$ . If  $\underline{c} \in S$  and  $\underline{u} \in \mathbf{R}^n$ , and we have  $\underline{c} + \underline{u} \in S$ , ( $\underline{u} \neq 0$ ), a *problem* could be to study the change in  $f(x)$  as  $x$  goes from  $\underline{c}$  to  $\underline{c} + \underline{u}$  along the line *segment*  $\underline{x} = \underline{c} + t\underline{u}$ , where  $t \in [0, 1]$ .

### 1.1. Directional Derivative

Study  $f$  as it *changes* in the direction of  $\underline{u}$ . We need a fairly general  $S$ , since the domains of functions can be complicated. For example, with  $f(x, y) = 1/\sqrt{x^2 + y^2 - 1}$ , if  $x, y \in S$ , then  $S = \{(x, y) \mid x^2 + y^2 \neq 1\}$ . Similarly, for  $f(x, y) = 1/(x^2 + y^2 - 1)^{1/2}$ , we need a *suitable*  $S$ . We usually take  $\underline{u}$  to be small. In the above,  $S$  is *everything* but the circle  $x^2 + y^2 = 1$ . If we take  $\underline{u}$  to be “big”, then we may not have  $\underline{c} + t\underline{u} \in S$  for all  $t \in [0, 1]$ .

Usually, we want  $S$  to be “**open**”. Suppose that  $r > 0$ . Let  $B(\underline{c}; r) = \{\underline{x} \mid \|\underline{x} - \underline{c}\| < r\}$  be the open ball with centre  $\underline{c}$ , radius  $r$ . *Definition*:  $S \subseteq \mathbf{R}^m$  is open if for all  $\underline{c} \in S$ , there is *some*  $r > 0$  with  $B(\underline{c}; r) \subseteq S$ . If  $S$  is *open* and  $t$  is small enough, then  $\underline{c} + t\underline{u}$  will be in  $B(\underline{c}; r)$ , because  $\|(\underline{c} + t\underline{u}) - \underline{c}\| = t\|\underline{u}\|$ , which can be as *small* as it needs to be (it will be  $< r$  if  $t < r/\|\underline{u}\|$ ).



**Definition** (for  $n = 1$ ): The *directional derivative* of  $f$  at  $\underline{c}$ , in the direction of  $\underline{u}$ , denoted by  $f'(\underline{c}; \underline{u})$ , is defined by  $f'(\underline{c}; \underline{u}) = \lim_{t \rightarrow 0} (f(\underline{c} + t\underline{u}) - f(\underline{c}))/t$ , wherever the *limit* exists. Pick a **co-ordinate** system. The partial derivatives are (directional derivatives in the *direction* of the axes)  $\frac{\partial f}{\partial x_k} \Big|_c = f'(\underline{c}; \underline{e}_k)$ , where  $\underline{e}_k$  is the **unit** vector  $(0, \dots, 0, 1, 0, \dots, 0)$ , with 1 in the  $k^{\text{th}}$  position.

**Example**:  $m = 2, n = 1, f(x, y, z) = xy + yz^2$  with  $S = \mathbf{R}^3$ .  $\frac{\partial f}{\partial x} = y, \frac{\partial f}{\partial y} = x + z^2$  and  $\frac{\partial f}{\partial z} = 2yz$ . (Taken at a *general* point  $\underline{c} = (x, y, z)$ ). **Example**:  $\underline{a} = (a_1, \dots, a_m) \in \mathbf{R}^m, f(\underline{x}) = \underline{a} \cdot \underline{x}$ . Find  $\delta f / \delta x_k$  from *first* principles. Now  $f(\underline{x} + t\underline{e}_k) = \underline{a} \cdot (\underline{x} + t\underline{e}_k) = \underline{a} \cdot \underline{x} + t\underline{a} \cdot \underline{e}_k = \underline{a} \cdot \underline{x} + ta_k$ . Now  $\lim_{t \rightarrow 0} (f(\underline{x} + t\underline{e}_k) - f(\underline{x}))/t = \lim_{t \rightarrow 0} (\underline{a} \cdot \underline{x} + ta_k - \underline{a} \cdot \underline{x})/t = a_k$ . In **general**, for  $f'(\underline{x}; \underline{u})$ ,  $f(\underline{x} + t\underline{u}) = \underline{a} \cdot \underline{x} + t\underline{a} \cdot \underline{u} = \underline{a} \cdot \underline{x} + t \sum a_i u_i$ .  $f'(\underline{x}; \underline{u}) = \sum a_i u_i = \underline{a} \cdot \underline{u} = \nabla f \cdot \underline{u}$  (where  $\nabla f = (\delta f / \delta x_1, \dots, \delta f / \delta x_m)$ ).

**Theorem:** Suppose  $f: S \rightarrow \mathbf{R}$ , and assume all *partial* derivatives exist **everywhere** in  $S$ . Set  $\nabla f = (\delta f / \delta x_1, \dots, \delta f / \delta x_m) = \sum_{i=1}^m \delta f / \delta x_i \mathbf{e}_i$ , the *gradient* of  $f$ . Then for all  $\underline{c}$  and  $\underline{u}$ ,  $f'(\underline{c}; \underline{u})$  exists, and  $f'(\underline{c}; \underline{u}) = \nabla f \cdot \underline{u}$ .

## Tutorial

Remember in **partial** differentiation, only differentiate in the *order* indicated, and only w.r.t. the variable(s) *involved*. Rules for stationary points (Revision): If  $z = f(x, y)$ , then to test for **stationary** points, solve  $\delta z / \delta x = \delta z / \delta y = 0$ . (**Important**). Then let  $A = \delta^2 z / \delta x^2$ ,  $B = \delta^2 z / \delta x \delta y$  and  $C = \delta^2 z / \delta y^2$ . Form the **Hessian** matrix of  $f$ ,  $H = \begin{pmatrix} A & B \\ B & C \end{pmatrix}$ . If  $H$  is *positive* definite at a stationary point  $P$ , then it is a local min. If it is *positive negative*,  $P$  is a local max. If it's **indefinite**, then  $P$  is a saddle point. If  $\det(H)$  is zero, then the *test* fails.

So if  $AC > B^2$  and  $A < 0$ , we have a local max; if  $AC > B^2$  and  $A > 0$ , we have a local min; if  $AC < B^2$ , it is a saddle point; and if  $AC = B^2$ , the test fails. The **tangent** plane at the point  $P(x_0, y_0, z_0)$  on the surface  $f(x, y, z) = c$  is the **plane**  $f_x(P)(x-x_0) + f_y(P)(y-y_0) + f_z(P)(z-z_0) = 0$ . The **normal** line of the surface at  $P$  is the line  $x = x_0 + f_x(P)t$ ,  $y = y_0 + f_y(P)t$ ,  $z = z_0 + f_z(P)t$ .

**Convex** functions. A function  $f(x_1, \dots, x_n)$  is a convex function on a *convex* set  $S$  (in  $\mathbf{R}^n$ ) if for any  $x' \in S$  and  $x'' \in S$ ,  $f(cx' + (1-c)x'') \leq cf(x') + (1-c)f(x'')$  holds for  $0 \leq c \leq 1$ . Define the **Hessian**,  $H$ , of  $f(x_1, \dots, x_n)$ , as being the  $n \times n$  matrix whose  $ij^{\text{th}}$  entry is  $\delta^2 f / \delta x_i \delta x_j$ . Winston quotes the **following** theorem: Suppose that  $f(x_1, \dots, x_n)$  has continuous second-order partial *derivatives* for each point  $x = (x_1, \dots, x_n) \in S$ . Then  $f(x_1, \dots, x_n)$  is a convex function on  $S$  if and only if for each  $x \in S$ , all **principal minors** of  $H$  are non-negative.

**Example:** Find the equation of the *tangent* plane and the normal line to the quadric surface  $3x^2 + 2y^2 + z^2 + 4x + 5y + 6z = 3$  at  $P(-1, -1, 1)$ . A: Let  $f(x, y, z)$  be  $3x^2 + 2y^2 + z^2 + 4x + 5y + 6z - 3 = 0$ . Let  $P$  be the **point**  $(-1, -1, 1)$ .  $F_x(P) = 6x + 4 = 6(-1) + 4 = -2$ .  $F_y(P) = 4y + 5 = 4(-1) + 5 = 1$ .  $F_z(P) = 2z + 6 = 2(1) + 6 = 8$ . The *tangent* plane is  $f_x(P)(x-x_0) + f_y(P)(y-y_0) + f_z(P)(z-z_0) = 0$ ;  $-2(x-(-1)) + 1(y-(-1)) + 8(z-1) = 0$ ;  $-2x - 2 + y + 1 + 8z - 8 = 0$ ;  $-2x + y + 8z = 9$ . The line **normal** to the *surface* at  $P$  is given by  $x = x_0 + f_x(P)t$ ;  $x = -1 - 2t$ .  $y = y_0 + f_y(P)t$ ;  $y = -1 + t$ .  $z = z_0 + f_z(P)t$ ;  $z = 1 + 8t$ .

Q: Verify that the **following** function is convex on  $\mathbf{R}^3$ :  $f(x, y, z) = 3x^2 + 2y^2 + z^2 + 4x + 5y + 6z$ . A: First we need to *calculate* the Hessian Matrix with the entries shown in yellow.  $\delta f / \delta x = 6x + 4$ .

$\begin{bmatrix} 6 & 0 & 0 \\ 0 & 4 & 0 \\ 0 & 0 & 2 \end{bmatrix}$ 
 $\delta^2 f / \delta x^2 = 6$ . And  $\delta^2 f / \delta x \delta y = \delta^2 f / \delta x \delta z = 0$ .  $\delta^2 f / \delta y^2 = 4$ .  $\delta^2 f / \delta z^2 = 2$ . And  $\delta^2 f / \delta y \delta x = \delta^2 f / \delta y \delta z = 0$ .  $\delta^2 f / \delta z \delta x = \delta^2 f / \delta z \delta y = 0$ . Therefore,  $H$  = the **Green** Matrix.
  $\begin{bmatrix} \delta^2 f / \delta x^2 & \delta^2 f / \delta x \delta y & \delta^2 f / \delta x \delta z \\ \delta^2 f / \delta y \delta x & \delta^2 f / \delta y^2 & \delta^2 f / \delta y \delta z \\ \delta^2 f / \delta z \delta x & \delta^2 f / \delta z \delta y & \delta^2 f / \delta z^2 \end{bmatrix}$

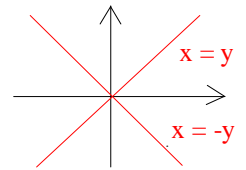
We now must test to see if all the *principle* minors of  $H$  are non-negative. If they are, then  $f(x, y, z)$  is a convex function. By deleting rows/columns 1 and 2, we obtain the **1st-order** principle minor 2, which is  $> 0$ . Similarly by deleting rows/columns 1 and 3, we get  $4 > 0$ , and by deleting rows/columns 2 and 3, we get  $6 > 0$ . By deleting row/column 1, we find the 2nd order principle minor,  $\det \begin{bmatrix} 4 & 0 \\ 0 & 2 \end{bmatrix} = 8 > 0$ . Row/column 2 gives  $\det \begin{bmatrix} 6 & 0 \\ 0 & 2 \end{bmatrix} = 12 > 0$ . And row/column 3 gives  $\det \begin{bmatrix} 6 & 0 \\ 0 & 4 \end{bmatrix} = 24 > 0$ . The **3rd-order principle minor** is the determinant of  $H$  itself, worked out expanding by *column 1* to be  $48 > 0$ . Since for all  $(x, y, z)$ , all principle minors of the Hessian are non-negative,  $f(x, y, z)$  **is** a convex function on  $\mathbf{R}^3$ .

## When do Directional Derivatives Exist?

If  $f'(c;u)$  exists in *every* direction, then the partial derivatives exist, but things go **wrong** for the converse. Limits of continuity in *many* variables (see T&F as well). Consider  $f: \mathbf{R}^2 \rightarrow \mathbf{R}$ , where  $f(x,y) = x+y$  if  $x = 0$  or  $y = 0$ , and 1 otherwise. Then  $\frac{\delta f}{\delta x}|_0 = 1$  and  $\frac{\delta f}{\delta y}|_0 = 1$ . But  $f(\underline{0}+t\underline{u})-f(\underline{0})/t$  for  $\underline{u} = (u_1, u_2)$  ( $u_1, u_2 \neq 0$ )  $= 1/t$ , which does not have a *limit* as  $t \rightarrow 0$ .

Consider  $S$  to be open connected,  $S \subseteq \mathbf{R}^2$ . Consider  $f: S \rightarrow \mathbf{R}^m$ . For  $m = 1$ ,  $\lim_{x \rightarrow x_0} f(x) = L$  means "to get values of  $f(x)$  near to  $L$ , we need only take *values* of  $x$  near  $x_0$ ", i.e. given  $\epsilon > 0$ , there exists a  $\delta > 0$  such that if  $|x-x_0| < \delta$ , then  $|f(x)-L| < \epsilon$ . Definition (in general)  $\lim_{x \rightarrow x_0} f(x) = L$  if given  $\epsilon > 0$ , there exists a  $\delta > 0$  such that if  $\|x-x_0\| < \delta$  and  $x \in S$ , then  $\|f(x)-L\| < \epsilon$ .

**Example:** consider  $f(x,y) = x^2-y^2/x^2+y^2$  for  $x \neq 0$ , and 0 for  $x = 0$ . Does  $\lim_{x \rightarrow 0} f(x)$  exist? (i) On the lines  $x = y$ ,  $x = -y$ ,  $f$  is **zero**. (ii) Along  $y = 0$ ,  $f(x,0) = 1$  if  $x \neq 0$  and 0 if  $x = 0$ . (iii) Along  $x = 0$ ,  $f(0,y) = -1$  if  $y \neq 0$ , and 0 if  $y = 0$ . Clearly, if  $L = \lim_{x \rightarrow x_0} f(x)$  exists, it **must** be at the same time 0, 1 and -1.



Try  $f(x,y) = 2x^2y/(x^4+y^2)$ ,  $(x,y) \neq (0,0)$  Does  $f(x)$  tend to *something* as  $\underline{x}$  tends to zero? Along  $x = 0$  and  $y = 0$ , the limit is zero. But along  $y = kx^2$  with non-zero  $k$ , work it out! Why does the method *work*? We want to see if there is an  $L$ , with  $f(\underline{x})$  tending to  $L$  as  $\underline{x}$  tends to  $\underline{x}_0$  (\*). A curve is a *continuous*  $x(t)$ ,  $t \in [0,r]$ , with  $\lim_{t \rightarrow 0} f(x(t)) = x_0$ . We **test** to see if  $\lim_{t \rightarrow 0} f(x(t))$  exists. If it doesn't, we stop, because (\*) cannot hold for *all*  $L$ .

If  $\lim_{x \rightarrow x_0} f(x) = L$ , then  $\lim_{t \rightarrow 0} f(x(t)) = L$  as well. If we **get**  $\lim_{t \rightarrow 0} f(x(t)) = L$  (i.e. the limit exists and is  $L$ ), then if  $\lim_{x \rightarrow x_0} f(x)$  exists, it *must* be  $L$ . But if  $x_1$  and  $x_2$  are two curves through  $x_0$ , and  $\lim_{t \rightarrow 0} f(x_1(t)) = L_1$  and  $\lim_{t \rightarrow 0} f(x_2(t)) = L_2$ ; if  $L_1$  is **not** equal to  $L_2$ , then  $\lim_{x \rightarrow x_0} f(x)$  *cannot* exist!

## Continuity

Suppose that  $S$  is an **open** subset of  $\mathbf{R}^m$ . Let  $\underline{x}_0 \in S$ ,  $f: S \rightarrow \mathbf{R}^n$ . Then we say that  $f$  is *continuous* at  $\underline{x}_0$  if  $\lim_{x \rightarrow x_0} f(x) = f(\underline{x}_0)$ . We say that  $f$  is continuous on  $S$  if it is continuous at *all points* of  $S$ . If  $f: S \rightarrow \mathbf{R}^n$ , then  $f(\underline{x}) = (f_1(\underline{x}), f_2(\underline{x}), \dots, f_n(\underline{x}))$ , and we can *specify*  $f$  by specifying the  $n$  **functions**  $f_1, \dots, f_n$  ( $S \rightarrow \mathbf{R}$ ).

**FACT:**  $f$  is continuous (cts) on  $S$  iff all the  $f_i$  ( $i = 1 \dots n$ ) are cts and *real-valued functions*. In particular, the  $k^{\text{th}}$  co-ordinate function,  $p_k: \mathbf{R}^n \rightarrow \mathbf{R}$ ,  $p_k(\underline{x}) = x_k$ , is cts (left as an exercise). Thinks: for  $n = 2$ , assume  $f_1$  and  $f_2$  are cts at  $\underline{x}_0$ . Given  $\epsilon > 0$ , we need to find a  $\delta > 0$  s.t. as *soon* as  $\|x-x_0\| < \delta$ , then we know that  $\|f(\underline{x})-f(\underline{x}_0)\| < \epsilon$ . Taking the end result to *pieces*, we want to make  $\|f(\underline{x})-f(\underline{x}_0)\| = ((f_1(\underline{x})-f_1(\underline{x}_0))^2 + (f_2(\underline{x})-f_2(\underline{x}_0))^2)^{1/2}$  nice and small.

Pick  $\delta_1 > 0$  so that if  $\|x-x_0\| < \delta_1$ , then  $\|f_1(x)-f_1(x_0)\| < k\epsilon$  ( $k$  is chosen later, and we can do this because  $f_1$  is cts). Pick  $\delta_2 > 0$ , so that if  $\|x-x_0\| < \delta_2$ , then  $\|f_2(x)-f_2(x_0)\| < k\epsilon$ . If  $\|x-x_0\| < \min(\delta_1, \delta_2)$ , then  $\|f(\underline{x})-f(\underline{x}_0)\| < ((k^2\epsilon^2) + (k^2\epsilon^2))^{1/2}$ . Pick  $k = 1/\sqrt{2}$  and  $\delta = \min(\delta_1, \delta_2)$ . Now carry on.

**FACTS:** (1) If  $f: S \rightarrow \mathbf{R}^n$ ,  $g: T \rightarrow \mathbf{R}$ , and  $T \subseteq \mathbf{R}^n$  are continuous, then so is  $gf: S \cap f^{-1}(T) \rightarrow \mathbf{R}^p$ .  
 $\text{dom}(gf) =$  the set of points for which  $gf(S)$  makes sense  $= \{s \mid s \in S \text{ and } f(S) \in T\}$  ( $s \in S$  so  $f(s)$  makes sense and  $f(S) \in T$  so  $gf(S)$  is defined). Recall the notation  $f^{-1}(T) = \{s \mid f(s) \in T\}$ . Note:  $f^{-1}(T)$  will be open if  $f$  is cts and  $T$  is open; and the intersection of 2 open sets is open. (2) If  $(\underline{x}_n)$  is a sequence of points in  $S$ , then  $\lim_{n \rightarrow \infty} \underline{x}_n$  exists, and is equal to  $a$ . If  $f$  is cts, then  $\lim_{n \rightarrow \infty} f(\underline{x}_n) = f(\underline{a})$ . (If this is the 1 variable case, replace  $|\dots|$  by  $\|\dots\|$ ).

**Back to differentiation.** Single variable case:  $f: S \rightarrow \mathbf{R}$ ,  $S \subseteq \mathbf{R}^1$ .  $S$  is differentiable at  $c \in S$  if  $\lim_{x \rightarrow c} \frac{f(x)-f(c)}{x-c}$  exists. ( $x \in S$ ). The value of the limit is  $f'(c)$ . For small  $(x-c)$ , we have small  $\frac{f(x)-f(c)}{x-c} - f'(c)$ . So  $f(x)-f(c) = f'(c)(x-c) +$  a small error, which tends to 0 as  $(x-c)$  tends to 0. So  $f(x)-f(c) = f'(c)(x-c) + e_c(x)(x-c)$ , and  $e_c(x) \rightarrow 0$  as  $x \rightarrow 0$ . Rewrite, with  $x = c+h$ , so that  $f(c+h) = f(c)+f'(c)h+E_c(h)|h|$ .

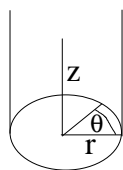
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Suppose that  $f: S \rightarrow \mathbf{R}^n$ , with  $S \subseteq \mathbf{R}^m$  (an open subset). Let  $\underline{c} \in S$ . Pick an  $r > 0$  such that  $B(\underline{c};r) \subseteq S$  (for safety). Then  $V \in \mathbf{R}^m$  such that  $\|v\| < r$ . **Definition:** We say that  $f$  is differentiable at  $\underline{c}$  if there is a linear transformation  $T: \mathbf{R}^m \rightarrow \mathbf{R}^n$ , such that  $f(\underline{c}+v) = f(\underline{c})+T_c(v)+\|v\|E_c(v)$ , where  $E_c(v) \rightarrow 0$  as  $v \rightarrow 0$ . (Given  $\epsilon > 0$ , there exists a  $\delta > 0$  such that if  $\|v\| < \delta$ , then  $E_c(v) < \epsilon$ ).  $T_c$  is called the total derivative of  $f$  at  $c$ . Look at  $v = t\underline{u}$  for a fixed direction  $\underline{u}$ :  $f(\underline{c}+t\underline{u}) = f(\underline{c})+T_c(t\underline{u})+|t|(\|\underline{u}\|)E_c(t\underline{u}) = f(\underline{c})+tT_c(\underline{u})+|t|(\|\underline{u}\|)E_c(t\underline{u})$ . So assuming  $n = 1$ ,  $\lim_{t \rightarrow 0} \frac{f(\underline{c}+t\underline{u})-f(\underline{c})}{t} = \lim_{t \rightarrow 0} (T_c(\underline{u}) \pm |t|\|\underline{u}\|E_c(t\underline{u})) = T_c(\underline{u})+0$ . But the LHS is  $f'(\underline{c};\underline{u})$ , the directional derivative of  $f$  at  $\underline{c}$  in the direction of  $\underline{u}$ .

**Recall:** If  $L: \mathbf{R}^m \rightarrow \mathbf{R}^n$ , pick the usual bases  $(e_k) = (0, \dots, 1, \dots, 0)$  (1 in the  $k^{\text{th}}$  place) in  $\mathbf{R}^m$  and  $\mathbf{R}^n$ , so that we can represent  $L$  as a matrix  $\underline{v} \in \mathbf{R}^{n \times m}$ , with  $\underline{v} = \sum_{j=1}^m v_j e_j^m$ . Therefore,  $\underline{v} = (v_1, \dots, v_m)$ . Now  $L(\underline{v}) = L(\sum_{j=1}^m v_j e_j^m) = \sum_{j=1}^m v_j L(e_j^m)$ . But  $L(e_j^m) = \sum_{k=1}^n l_{kj} e_k^n$ . (The matrix  $(l_{kj})$ , where this holds). Going back to  $f: S \rightarrow \mathbf{R}^n$ ,  $T_c(e_j^m) = (f_1(\underline{c}; e_j^m), f_2(\underline{c}; e_j^m), \dots, f_n(\underline{c}; e_j^m)) = ((\delta f_1 / \delta x_1)_c, \dots, (\delta f_n / \delta x_j)_c)$ . If  $f$  is differentiable at  $c$ , then w.r.t. the usual bases,  $T_c$  has matrix  $(\delta f_i / \delta x_j)$ , i.e. the Jacobian matrix of  $f$  (at  $c$ ). If  $f: S \rightarrow \mathbf{R}^n$  ( $S$  in  $\mathbf{R}^m$ ), the  $T_c$  is given by  $T_c(\underline{u}) = f'(c; \underline{u}) = \nabla f(c) : \underline{u} = (\delta f / \delta x_1, \dots, \delta f / \delta x_m)$ .

➤ 12th October 1999

**Example:** Find the derivative of  $f$  at  $c$  (in the direction of  $\underline{u}$ ), where  $f(x,y,z) = x^3 - xy^2 - z$ ,  $\underline{u} = (2, -3, 6)$  &  $\underline{c} = (1, 1, 0)$ .  $\nabla f = (3x^2 - y^2, -2xy, -1)$ . The direction of  $\underline{u}$  is  $\underline{u} / \|\underline{u}\|$ .



$\|\underline{u}\| = \sqrt{49} = 7$ , so  $\underline{u} = (2/7, -3/7, 6/7)$ .  $\nabla f(c) = (2, -2, 1)$ ;  $f'(c; \underline{u}_{\text{normalised}}) = \begin{bmatrix} \cos \theta & -r \sin \theta & 0 \\ \sin \theta & r \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix} (2, -1, 1) \cdot (2/7, -3/7, 6/7) = 4/7 + 6/7 + 6/7$ . **Example:** Let  $f: \mathbf{R}^3 \rightarrow \mathbf{R}^3$ ;  $f(r, \theta, z) = (r \cos \theta, r \sin \theta, z)$ . The Jacobian matrix is as shown in yellow.

## The Chain Rule

Let  $f$  and  $g$  be functions. Assume  $h = f \circ g$  is defined in a neighbourhood of a point  $\underline{a}$ . Assume  $g$  is differentiable at  $\underline{a}$ , with total derivative  $g'(\underline{a})$  (a linear transformation). Set  $\underline{b} = g(\underline{a})$ . Assume  $f$  is differentiable at  $\underline{b}$  with total derivative  $f'(\underline{b})$  (a linear transformation). Then  $h$  is differentiable at  $\underline{a}$ ; its total derivative is  $f'(\underline{b}) \circ g'(\underline{a})$ , the composite linear transformation. If we fix co-ordinates, and write  $Df(\underline{b})$  for the Jacobian matrix of  $f$  at  $\underline{b}$ , etc., we get  $Df(\underline{a}) = Df(\underline{b}) \circ Dg(\underline{a})$ .

**Example:**  $g: \mathbf{R} \rightarrow \mathbf{R}^2$ ,  $f: \mathbf{R}^2 \rightarrow \mathbf{R}$ .  $g(t) = (x(t), y(t))$  in  $\mathbf{R}^2$ .  $f(x,y)$  in  $\mathbf{R}$ .  $D_g(\underline{a}) = \begin{pmatrix} dx/dt \\ dy/dt \end{pmatrix}_a$  (a column vector).  $D_f(\underline{b}) = \begin{pmatrix} \delta f / \delta x \\ \delta f / \delta y \end{pmatrix}$ . So  $h(t) = f(x(t), y(t))$ ,  $dh/dt = \delta f / \delta x \delta x / \delta t + \delta f / \delta y \delta y / \delta t$ .

## Tutorial Exercises

**Q:** For the following, give the **domain** (open or closed?); whether the function has a limit at  $(0,0)$ ; and decide whether the function could be made *continuous* at the origin. (a)  $y^2/(x^2+y^2)$ . (b)  $(x-y)/(x+y)$ . (c)  $(x^3+y^3)/(x^2+y^2)$ . (d)  $(x+y)/(x^2+y^2)$ . (e)  $[\sin(x+y)]/(x+y)$ . **A:** (a) Makes **sense** provided  $x^2+y^2 \neq 0$ . So the domain is  $D_f = \mathbf{R}^2 \setminus \{(0,0)\}$ , which is an *open* set. The function cannot have a limit as  $(x,y) \rightarrow (0,0)$ , since going to  $(0,0)$  along the line  $x = y$  gives a limit of a  $1/2$ , and going along  $x = 2y$  gives a limit of  $1/5$ . If it **did** have a limit  $l$ , that limit would be the answer along any path through the origin. In addition, the function *cannot* be made cts at the origin.

(b) Causes problems where  $x = -y$ . So  $D_f = \{(x,y) \mid x \neq -y\}$ . This is an **open** set (visualise the drawing). Along  $x = 0$ , the function has constant value  $-1$ . Along  $y = 0$ , it has value  $1$ . So **there is no** limit; and it cannot be made continuous at the origin.

(c) Goes wrong at  $x^2+y^2 = 0$  i.e. at  $(0,0)$  only. So  $D_f = \mathbf{R}^2 \setminus \{(0,0)\}$ , open. Trying different paths and getting  $0$  suggests that the limit *should* be zero. Recall the definition of  $\lim_{(x,y) \rightarrow (a,b)} f(x,y) = L$ : Given  $\epsilon > 0$ , there *exists* a  $\delta > 0$  s.t.  $\|(x,y)-(a,b)\| < \delta \Rightarrow |f(x,y)-L| < \epsilon$ . Convert the expression for  $f$  to polar **co-ordinates**. Using  $x = r\cos\theta$  and  $y = r\sin\theta$ ,  $f(x,y)$  becomes  $r^3(\cos^3\theta + \sin^3\theta)/r^2$ . (As  $\|(x,y)\| = r$ , we want to make this small when  $r$  is small). Given  $\epsilon > 0$ , take  $\delta = \epsilon/2$ , then  $|f(x,y) - 0| = r|\cos^3\theta + \sin^3\theta| < 2r$ . If  $\|(x,y)\| < \delta$ , then  $|f(x,y)| < 2\|(x,y)\| \leq \epsilon$ , i.e.  $\lim_{(x,y) \rightarrow 0} f(x,y) = 0$ . Set  $f(0,0) = 0$  to get a *continuous* function anywhere on  $\mathbf{R}^2$ .

(d) *Answer:*  $D_f = \mathbf{R}^2 \setminus \{(0,0)\}$ ; the limit does not exist as different paths give different “limits” — most tend to **infinity**. (e) As  $\sin(u)/u \rightarrow 0$  as  $u \rightarrow 0$ , this has a *removable* discontinuity — set  $f(0,0) = 0$  and you get a continuous function on  $\mathbf{R}^2$ .

**Q:** Which of the **following** functions is continuous at the origin?: (a)  $f(x,y) = xy/(x^2+y^2)$  if  $(x,y) \neq (0,0)$ ; and  $f(x,y) = 0$  if  $(x,y) = (0,0)$ ; (b)  $f(x,y) = (x^3-y^3)/(x^2+y^2)$  if  $(x,y) \neq (0,0)$ , and  $f(x,y) = 0$  if  $(x,y) = (0,0)$ . **A:** (a) Along  $x = y$ , this gives a *limit* of a  $1/2$  as  $(x,y) \rightarrow (0,0)$  (along **other** curves, you get different limits). So as  $1/2 \neq 0$ ,  $f$  is not continuous at  $(0,0)$ . (b)  $\lim_{(x,y) \rightarrow (0,0)} f(x,y) = 0$  (similar to question 1c), so it is **continuous** at  $(0,0)$ .

**Q:** Determine whether or not the *following function* is continuous everywhere in  $\mathbf{R}^2$ :  $f(x,y) = (x^3-y^3)/(x-y)$  for  $x \neq y$ , and  $f(x,y) = 3x^2$  for  $x = y$ . **A:** If  $x \neq y$ ,  $f(x,y) = x^3-y^3/x-y = x^2+xy+y^2$ . This is cts *everywhere* in  $\mathbf{R}^2 \setminus \{(\alpha,\alpha) \mid \alpha \in \mathbf{R}\}$ , so we need only check *continuity* on this “diagonal”.  $\lim_{(x,y) \rightarrow (a,a)} f(x,y) = 3a^2$ . This is a claim. Sketch Proof: From the definition, we need to check that if given an  $\epsilon > 0$ , there exists a  $\delta > 0$ , etc. Consider the first few “*obvious*” lines, and then split to consider the **two** cases  $x = y$  and  $x \neq y$  separately.

**Q:** In the **discussion** of continuity of a function  $f$ , ( $f: S \rightarrow \mathbf{R}^n$ ), it says “ $f$  is cts if it is cts at all points of  $S$ ” and “ $f$  is cts at a *element*  $S$  if  $\lim_{x \rightarrow a} f(x) = f(a)$ . This *definition* of continuity uses the notion of a limit, so a reminder:  $\lim_{x \rightarrow a} f(x) = f(a)$  means: Given  $\epsilon > 0$ , there’s a  $\delta > 0$  s.t. if  $|x-a| < \delta$ , then  $|f(x)-f(a)| < \epsilon$ .

For a **different**  $a$ , we may need a different  $\delta$ . Example:  $f(x) = 1/x$  for  $x \in \mathbf{R} \setminus \{0\}$ . Fix  $\varepsilon = 10$  and test for *the continuity* of  $f$  at each of the points  $a_n = 1/n$ . ( $n \in \mathbf{N} \setminus \{0\}$ ). In each case, give a suitable  $\delta > 0$ . Can you use a single  $\delta$  for all  $a_n$ ? In general, functions that are just *continuous* can have awkward properties. The stronger notion of uniform continuity is therefore introduced. In this, the **SAME**  $\delta$  works for all points. Check that  $f(x) = x^2$  for  $x \in [-1, 1]$  is *uniformly* continuous.

**Q: Generalising** the above to two variables, (i) Prove first that the function  $f(x, y) = 5x + xy$  is cts at any  $(a, b)$  in  $\mathbf{R}^2$ ; (ii) Prove that the function above is *uniformly* continuous in the square region  $[0, 1] \times [0, 1]$ , that is where  $0 \leq x \leq 1$  and  $0 \leq y \leq 1$ .

**A:** Suppose we could **find** a  $\delta$  which worked for all the  $a_n$ . Any smaller  $\delta$  would therefore also work, so we can find **some**  $n$  with  $1/n < \delta$ , and we can assume that  $n > 100$ , say. Now **take**  $x = 1/n$ ,  $|x - a_{2n}| = |1/n - 1/2n| = 1/2n < \delta$ . But  $|f(x) - f(a_{2n})| = |n - 2n| = |n| > 100$ . So **NO WAY** can we get the same  $\delta$  everywhere. For  $f(x) = x^2$ , look at the *usual* direct proof of continuity. Given  $\varepsilon > 0$ , check the continuity at  $a \in [-1, 1]$ .  $|x^2 - a^2| = |(x-a)(x+a)| \leq 2|x-a|$  (since  $|x+a| \leq 2$  if  $x, a \in [-1, 1]$ ). So to get  $|x^2 - a^2| < \varepsilon$ , we need only take  $|x-a| < \varepsilon/2$  (which does *not* depend on  $a$ ) so we have:  $f$  is **uniformly** continuous *everywhere* in  $[-1, 1]$ .

(b) Try the **same** approach. We are given  $\varepsilon > 0$ . Check for *continuity* at  $(a, b) \in \mathbf{R}^2$ . Examine:  $|f(x, y) - f(a, b)| = |5(x-a) + (xy-ab)| \leq 5|x-a| + |xy-ab|$ . (*Triangle inequality*). (Note: if  $\|(x, y) - (a, b)\|$  is small, we can *make*  $5|x-a|$  small (it will be anyway!) so we have to **concentrate** on  $(xy-ab)$ ). Now  $|xy-ab| = |x(y-b) + b(x-a)| \leq |x||y-b| + |b||x-a|$ . (**Same** old trick — take it off and add it in again). If  $\|(x, y) - (a, b)\|$  is small, say  $< \delta_1$ , then  $|x-a| < \delta_1$ .

(\*) So  $|x| < |a| + \delta_1$ . Also,  $|y-b| < \delta_1$  so  $|xy-ab| \leq \delta_1(|a| + \delta_1) + |b|\delta_1$ . This can now be *made* smaller than  $\varepsilon/6$ , say, by picking  $\delta_1$  to be small enough (the *details* don't matter!) If we pick  $\delta$  to be smaller than this  $\delta_1$ , and also smaller than  $\varepsilon/6$ , we will get  $|f(x, y) - f(a, b)| < \varepsilon/6 + \varepsilon/6 = \varepsilon$ . (But  $\delta_1$  will depend on  $(a, b)$  in *general* here!).

(ii) Now if we *restrict our* attention to  $(x, y) \in [0, 1] \times [0, 1]$ , then in *line* (\*) above,  $|x| < 1$  and  $|b| < 1$ . So we **have**  $|xy-ab| \leq 2\delta_1$ , so we can choose  $\delta_1$  so that  $2\delta_1 < \varepsilon/6$  *independently* of  $(a, b)$ . (This is not the “**full**” elegant solution). Fact: **any** continuous function on a closed bounded subset of  $\mathbf{R}^m$  is *uniformly continuous*.

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(2) Let  $g(t) = (x(t), y(t), z(t))$ ;  $g: S \rightarrow \mathbf{R}^3$ ; (where  $S$  is a subset);  $f: \mathbf{R}^3 \rightarrow \mathbf{R}$ ;  $h$  is a composite:  $f(x(t), y(t), z(t)) = h(t)$ . Then  $dh/dt = \delta f/\delta x \cdot dx/dt + \delta f/\delta y \cdot dy/dt + \delta f/\delta z \cdot dz/dt$ .

(3) Let  $f: f(x, y, z)$ ;  $f: \mathbf{R}^3 \rightarrow \mathbf{R}$ ;  $x = x(r, s)$ ,  $y = y(r, s)$ ,  $z = z(r, s)$ . Write  $g(r, s) = (x(r, s), y(r, s), z(r, s))$ . Let  $g: \mathbf{R}^2 \rightarrow \mathbf{R}^3$ . Now look at the *definitions* for  $D_g(\underline{a})$  on the right and  $D_f(\underline{b}) = (\delta f/\delta x, \delta f/\delta y, \delta f/\delta z)$ . Let  $h = f \circ g: \mathbf{R}^2 \rightarrow \mathbf{R}$ . Then  $D_h(\underline{a}) = (\delta h/\delta r, \delta h/\delta s)$ ;  $D_h(\underline{a}) =$  see the left. So  $\delta h/\delta r = \delta f/\delta x \cdot \delta x/\delta r + \delta f/\delta y \cdot \delta y/\delta r + \delta f/\delta z \cdot \delta z/\delta r$ . Similarly for  $\delta h/\delta s$ .

$$D_g(\underline{a}) = \begin{bmatrix} \frac{\partial x}{\partial r} & \frac{\partial x}{\partial s} \\ \frac{\partial y}{\partial r} & \frac{\partial y}{\partial s} \\ \frac{\partial z}{\partial r} & \frac{\partial z}{\partial s} \end{bmatrix}$$

## Inverse Functions (1 Variable Case)

Let  $f$  be a function,  $f: S \rightarrow \mathbf{R}$ ,  $S \subseteq \mathbf{R}$  (open). Consider a  $c \in S$  such that  $f'(x)$  exists and is continuous in a neighbourhood of  $c$ . Assume that  $f'(c) \neq 0$ . Then if  $f'(c) > 0$ ,  $f$  is **monotonically** increasing near  $c$ . If  $f'(c) < 0$ ,  $f$  is monotonically *decreasing* near  $c$ . If  $x > c$  and  $f'(c) > 0$ , (we need  $f(x) \geq f(c)$ ), **look** at  $f(x) - f(c) = f'(c)(x - c) + \text{small error}(|x - c|)$ . (This small error *tends* to 0 as  $x \rightarrow c$ ). So as  $f'(c) > 0$  and  $x - c > 0$ ,  $f(x)$  will be greater than  $f(c)$ . (for  $x$  near enough to  $c$ ). Similarly for  $c > x$ . And **similarly** for  $f'(c) < 0$ : implies *monotonically decreasing*.

Thus near  $c$ ,  $f$  will have a (**locally** defined) inverse function. If  $y = f(x)$ , we can find a function  $g$  such that  $g(y) = x$ . Now  $f(x) - f(c) = f'(c)(x - c) + \text{small error}$ . So  $(x - c) = \frac{f(x) - f(c)}{f'(c)} + \text{small error}$ . Write  $y = f(x)$  and  $b = f(c)$ , so that  $g(y) = x$ ;  $g(b) = c$ ;  $g(y) - g(b) = \frac{1}{f'(c)}(y - b) + \text{small error}$ . So  $g'(b)$  exists and equals  $\frac{1}{f'(c)}$  (**Detailed** argument omitted).

**Note:** As  $g$  is a (*local*) inverse for  $f$  near  $c$ ,  $f \cdot g = \text{Id}$  near  $f(c) = b$ , and  $g \cdot f = \text{Id}$  near  $c$ . The **chain** rule gives  $g'(b) = \frac{1}{f'(c)}$  anyway. ( $I = D + D_g$ ).

➤ 19th October 1999

Can we **generalise** the single variable case to *the many* variable case? Suppose that  $S \subseteq \mathbf{R}^m$  (open);  $f = (f_1, \dots, f_n)$ ;  $S \rightarrow \mathbf{R}^n$ ;  $\underline{a} \in S$ ; and all  $\delta f_i / \delta x_j$  are continuous on some *open set* containing  $\underline{a}$ . Then  $f(\underline{x}) - f(\underline{a}) = f'(\underline{a})(\underline{x} - \underline{a}) + \text{small error}$ . (Note:  $f'(\underline{a}) = D_f(\underline{a})$ ). We write  $\underline{y} = f(\underline{x})$ ;  $\underline{b} = f(\underline{a})$  as *before*. Then  $\underline{y} - \underline{b} = D_f(\underline{a})(\underline{x} - \underline{a}) + \text{small error} (*)$ .

Think of it as  $(\underline{y} - \underline{b}) = T(\underline{x} - \underline{a})$ , with  $T$  *linear*. We can solve it **provided**  $\det(T) \neq 0$ , i.e. as long as  $T$  is invertible. Oops! — we will need  $n = m$ !. We therefore can hope to *solve* (\*) for  $\underline{x}$  in terms of  $\underline{y}$  provided that  $\det D_f(\underline{a}) \neq 0$ , and then  $\underline{x}$  will be **determined** by  $\underline{y}$  locally near  $\underline{y} = \underline{b}$ .

**Notation:** The *Jacobian* determinant is  $J_f(\underline{a}) = \det(D_f(\underline{a}))$ . Could also **use**  $J_f(\underline{x}) = \det(D_f(\underline{x}))$ .  $J_f(\underline{x})$  will be a sum of *products* of the  $(\delta f_i / \delta x_j)(\underline{x})$ 's. Hence  $J_f(\underline{x})$  is a **continuous** function of  $x$  on some *open* neighbourhood of  $\underline{a}$ . If  $J_f(\underline{a}) \neq 0$ , then  $J_f(\underline{x}) \neq 0$  in a **neighbourhood** around  $\underline{a}$ .

**Consequence:**  $f$  must be 1-to-1 near  $\underline{a}$ . Justification: because  $f(\underline{x}) = f(\underline{x}')$ ;  $0 = f(\underline{x}) - f(\underline{x}') = D_f(\underline{c})(\underline{x} - \underline{x}')$ , [Using a **many** variable version of the *Mean Value Theorem* — details not given], where  $\underline{c}$  lies **between**  $\underline{x}$  and  $\underline{x}'$ . So  $D_f(\underline{c})$  will be *invertible* provided all of  $\underline{x}$ ,  $\underline{x}'$  and  $\underline{c}$  are near **enough** to  $\underline{a}$ . But then  $0 = D_f(\underline{c})(\underline{x} - \underline{x}')$ , and so  $\underline{x} - \underline{x}' = 0$ , i.e.  $\underline{x} = \underline{x}'$ . Then we **expect** to have a (*locally* defined) inverse  $g$  of  $f$  near  $\underline{a}$ :  $D_g(\underline{b}) = D_f(\underline{a})^{-1}$ , where  $\underline{b} = f(\underline{a})$ . **Note:** the *notation*  $\delta(f_1, \dots, f_m) / \delta(x_1, \dots, x_m)$  is sometimes used for  $J_f(\underline{x})$ .

## Curvilinear Co-ordinates

We will look at *Polar co-ordinates* in  $\mathbf{R}^2$ , *cylindrical co-ordinates* in  $\mathbf{R}^3$ , and *spherical co-ordinates* in  $\mathbf{R}^3$ . **Polar Co-ordinates.**  $(x, y) \rightarrow (r, \theta)$  [cartesian  $\rightarrow$  polar]. Here,  $x = r \cos \theta$  and  $y = r \sin \theta$ . Or,  $r = \sqrt{x^2 + y^2}$  with  $r > 0$ , and  $\theta = \tan^{-1}(y/x)$  with  $0 < \theta < \pi$ .

## Workshop 3

**Recall** that a function  $f(x_1, \dots, x_n)$  is a convex function on a *convex set*  $S$  (in  $\mathbf{R}^n$ ) if for any  $x' \in S$  and  $x'' \in S$ ,  $f(cx' + (1-c)x'') \leq cf(x') + (1-c)f(x'')$  holds for  $0 \leq c \leq 1$ . Also, a **subset**  $X \subseteq \mathbf{R}^n$  is convex if for any *pair of points*  $\underline{x}, \underline{y} \in X$  and any  $t$  with  $0 \leq t \leq 1$ , then  $t\underline{x} + (1-t)\underline{y} \in X$  (Thus  $X$  is **convex** if the line segment joining any *two points of*  $X$  is completely within  $X$ ).

Recall that the **Hessian**,  $H$ , of  $f(x_1, \dots, x_n)$ , is the  $n \times n$  matrix whose  $ij$ th entry is  $\delta^2 f / \delta x_i \delta x_j$ .  $H(\underline{c})$  is the Hessian **evaluated** at a point  $\underline{c}$ . With this, **Taylor's** theorem implies that  $f(\underline{x}) = f(\underline{x}_0) + D_f(\underline{x}_0) \cdot (\underline{x} - \underline{x}_0) + \frac{1}{2!} (\underline{x} - \underline{x}_0)^T H(\underline{c}) (\underline{x} - \underline{x}_0)$  for some  $\underline{c}$  *between*  $\underline{x}$  and  $\underline{x}_0$ . (Note: **additional** details such as differentiability, etc. on  $X$ ,  $X$  being convex are needed. These are *omitted*).

**Winston** has the following theorem: (assuming  $X$  is *convex*): Suppose  $f(x_1, \dots, x_n)$  has continuous second-order partial derivatives for each **point**  $x = (x_1, \dots, x_n) \in X$ . Then  $f(x_1, \dots, x_n)$  is a convex function on  $X$  iff for each  $x \in X$ , all **principal minors** of  $H$  are non-negative.

This condition on the *principal minors* is more simply stated using the notion of a +ve **(semi) definite** matrix: A *real symmetric matrix*  $A$  is positive semi definite if  $\underline{x}^T A \underline{x} \geq 0$ ; it is positive **definite** if  $\underline{x}^T A \underline{x} > 0$ . With this, the above **result** states: Suppose  $f(x_1, \dots, x_n)$  has continuous second-order partial derivatives for each *point*  $x = (x_1, \dots, x_n) \in X$ . Then  $f(x_1, \dots, x_n)$  is a convex function on  $X$  iff for each  $\underline{c} \in X$ ,  $H(\underline{c})$  is +ve semi definite.

**Q:** Suppose that  $X$  is a **convex** subset of  $\mathbf{R}^1$  (is an interval) and that we have the function  $f: X \rightarrow \mathbf{R}^1$ . Suppose that  $f''(x)$  exists for all  $x \in X$ . Prove that  $f$  is a *convex function* iff  $f''(x) \geq 0$  for all  $x \in X$ . Hints: look at *both* sides; use the **one** variable form of the Taylor expansion. Fiddle using  $f''(x) \geq 0$ .

**Q:** Determine whether *each of the following functions* is convex, concave or neither: (a)  $f(x) = x^3$ ,  $x \geq 0$ ; (b)  $f(x) = x^3$ ,  $x \in \mathbf{R}$ ; (c)  $f(x) = 1/x$ ,  $x > 0$ ; (d)  $f(x) = x^a$ , ( $0 \leq a \leq 1$ ),  $x > 0$ .

**Q:** Calculate the **Hessian**,  $H(\underline{c})$ , of  $f(x, y) = x^4 + 2xy^2 + y^2$ . Find a **region** of the plane on which  $H(\underline{c})$  is +ve *semidefinite* everywhere. (ii) Look at the above proof. Does it generalise to higher dimensions? Try to **prove** the previous stated theorem.

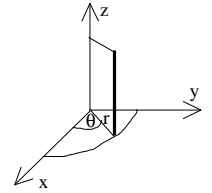
➤ 25th October 1999

## Curvilinear Co-ordinates

**Set**  $g: S \rightarrow \mathbf{R}^2$ ;  $S = \{(r, \theta) \mid 0 < r, 0 < \theta < 2\pi\}$ ;  $(r, \theta) \rightarrow (r \cos \theta, r \sin \theta)$ . [So  $f: \mathbf{R}^2 \rightarrow \mathbf{R}$ , say in  $x, y$  terms, becomes  $f \circ g$  in terms of  $(r, \theta)$ ;  $g$  is the “*substitution*” for  $(x, y)$  in terms of  $(r, \theta)$ ]. Note:  $D_g(r, \theta) = \begin{pmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \end{pmatrix}$ . All  $\delta g / \delta r$  and  $\delta g / \delta \theta$  are *continuous*. The Jacobian is  $J_g(r, \theta) = \det(D_g(r, \theta)) = r > 0$  (where the  $r$  comes from  $r \cos^2 \theta + r \sin^2 \theta$ ). So  $g$  is **locally** invertible.

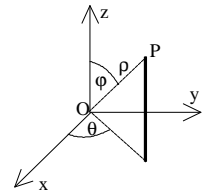
## Cylindrical Co-ordinates

We represent a *point* by  $(r, \theta, z)$ , where (1)  $(r, \theta)$  is the expression for the *vertical* projection in polar co-ordinates; (2)  $z$  is the (old) **vertical** co-ordinate. Here,  $x = r\cos\theta$ ,  $y = r\sin\theta$ , and  $z = z$ . Further,  $r = \sqrt{(x^2+y^2)}$ ;  $\theta = \tan^{-1}(y/x)$ ; and  $z = z$ .



## Spherical Polar Co-ordinates

These are *represented* by  $(\rho, \phi, \theta)$ , where  $\rho$  is the *distance* to P from O;  $\phi$  is the *distance* between the line OP and the **positive** z-axis; and  $\theta$  is as in cylindrical co-ordinates. Here,  $x = \rho\sin\phi\cos\theta$ ,  $y = \rho\sin\phi\sin\theta$ , and  $z = \rho\cos\phi$ . (**These** come from  $r, \theta, z$ ). Also,  $r = \rho\sin\phi$ ;  $\theta = \theta$ ; and  $z = \rho\cos\phi$ . (**These** come from  $\rho, \phi, \theta$ .) And  $\rho = \sqrt{(r^2+z^2)} = \sqrt{(x^2+y^2+z^2)}$ . Therefore,  $x = \rho\sin\phi\cos\theta$ ;  $y = \rho\sin\phi\sin\theta$ ; and  $z = \rho\cos\phi$ .



## An Example on the Use of Change in Co-ordinates

Zaplace's equation in two dimensions is  $\nabla^2 f = 0$ , i.e.  $\frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2} = 0$ . What does this look like in *polars*? Now  $\frac{\partial f}{\partial x} = \frac{\partial f}{\partial r} \frac{\partial r}{\partial x} + \frac{\partial f}{\partial \theta} \frac{\partial \theta}{\partial x}$ , and *similarly* for  $\frac{\partial f}{\partial y}$ . Differentiate  $x^2+y^2 = r^2$  w.r.t.  $x$ :  $2x = 2r \frac{\partial r}{\partial x}$ , so  $\frac{\partial r}{\partial x} = \frac{x}{r} = \cos\theta$ . Also,  $\frac{\partial r}{\partial y} = \sin\theta$ . Now  $y/x = \tan\theta$  implies  $\sec^2\theta \frac{\partial \theta}{\partial x} = -\frac{y}{x^2}$ . Or  $\frac{\partial \theta}{\partial x} = -\frac{y}{x^2} \cos^2\theta = -\frac{\sin\theta}{r}$ . *Similarly*,  $\frac{\partial \theta}{\partial y} = \frac{\cos\theta}{r}$ , so that  $\frac{\partial f}{\partial x} = \cos\theta \frac{\partial f}{\partial r} - \frac{\sin\theta}{r} \frac{\partial f}{\partial \theta}$ .

Since  $f$  is *arbitrary*,  $\frac{\partial}{\partial x} = \cos\theta \frac{\partial}{\partial r} - \frac{\sin\theta}{r} \frac{\partial}{\partial \theta}$ . So  $\frac{\partial^2 f}{\partial x^2} = ((\cos\theta \frac{\partial}{\partial r} - \frac{\sin\theta}{r} \frac{\partial}{\partial \theta}) | \cos\theta \frac{\partial f}{\partial r} - \frac{\sin\theta}{r} \frac{\partial f}{\partial \theta}) = \cos\theta \frac{\partial}{\partial r} (\cos\theta \frac{\partial f}{\partial r}) + \cos\theta \frac{\partial}{\partial r} (-\frac{\sin\theta}{r} \frac{\partial f}{\partial \theta}) - \frac{\sin\theta}{r} \frac{\partial}{\partial \theta} (\cos\theta \frac{\partial f}{\partial r}) - \frac{\sin\theta}{r} \frac{\partial}{\partial \theta} (-\frac{\sin\theta}{r} \frac{\partial f}{\partial \theta}) = \cos^2\theta \frac{\partial^2 f}{\partial r^2} - \sin\theta \cos\theta [-\frac{1}{r^2} \frac{\partial f}{\partial \theta} + \frac{1}{r} \frac{\partial^2 f}{\partial r \partial \theta}] + \dots = \cos^2\theta \frac{\partial^2 f}{\partial r^2} + \frac{\sin^2\theta}{r^2} \frac{\partial^2 f}{\partial \theta^2} - \frac{2}{r} \sin\theta \cos\theta \frac{\partial^2 f}{\partial r \partial \theta} + \frac{2\sin\theta}{r^2} \cos\theta \frac{\partial f}{\partial \theta} + \frac{\sin^2\theta}{r} \frac{\partial f}{\partial r}$ .

*Similarly* for  $\frac{\partial^2 f}{\partial y^2}$ , and then we use  $\sin^2\theta + \cos^2\theta = 1$ . This gives  $\nabla^2 f = \frac{\partial^2 f}{\partial r^2} + \frac{1}{r^2} \frac{\partial^2 f}{\partial \theta^2} + \frac{1}{r} \frac{\partial f}{\partial r}$ . As an application, if you are modelling a *phenomenon* which has circular symmetry, then  $f$  can be assumed to be independent of  $\theta$ . So  $0 = \nabla^2 f = \frac{d^2 f}{dr^2} + \frac{1}{r} \frac{df}{dr}$ ;  $r \frac{d^2 f}{dr^2} + \frac{df}{dr} = 0$ , or  $\frac{d}{dr}(r \frac{df}{dr}) = 0$ . So  $r \frac{df}{dr} = A$ , a *constant*;  $\frac{df}{dr} = \frac{A}{r}$ , so  $f(r) = A \ln(r) + B$  (where **B** is *another* constant).

➤ 26th October 1999

Two *useful things* left out of the syllabus: (1) **Taylor Series**:  $f(x) = f(x_0) + \frac{df}{dx}(x_0)(x-x_0) + \frac{1}{2!} \frac{d^2 f}{dx^2}(c)(x-x_0)^2$ , where the last term with  $c$  in it is the *error* term. **Generalising** to  $n$ -dimensions,  $f(\underline{x}) = f(\underline{x}_0) + D_f(\underline{x}_0)(\underline{x}-\underline{x}_0) + \frac{1}{2}(\underline{x}-\underline{x}_0)^T H(c)(\underline{x}-\underline{x}_0)$ , for some  $c$  lying *between*  $\underline{x}$  and  $\underline{x}_0$ . (2) **Implicit Function Theorem**: Involves e.g.  $x^2+y^2 = 1$ ;  $y = \pm\sqrt{(1-x^2)}$ .

## Assignment 1 (Not Assessed)

Q: Find the *equation* of the tangent plane and the normal line to the quadric surface  $3x^2+2y^2+z^2+4x+5y+6z = 3$  at the **point**  $P(-1,-1,1)$ . A:  $f_x = \frac{\partial f}{\partial x} = 6x+4$ ;  $f_y = 4y+5$ ;  $f_z = 2z+6$ . At  $P$ ,  $f_x(P) = -2$ ;  $f_y(P) = 1$ ; and  $f_z(P) = 8$ . So the **tangent** plane is  $-2(x+1)+(y+1)+8(z-1) = 0$  or  $-2x+y+8z-9 = 0$ . (Note the use of  $\nabla f(P)$  to get  $\nabla f(P) \cdot (\underline{x}-\underline{x}_0) = 0$ . Geometrically,  $\nabla f(P)$  is a *normal direction* to the surface. This gives: ...). The **normal** line to the surface at  $P$  is  $\underline{x} = \underline{x}_0 + \nabla f(P)t$ , i.e.  $x = -1-2t$ ,  $y = -1+t$ ;  $z = 1+8t$ .

Q: Verify that the function  $f(x,y,z) = 3x^2+2y^2+z^2+4x+5y+6z$  is *convex* on  $\mathbf{R}^3$ . A: Same  $f$  as above.  $\frac{\partial^2 f}{\partial x^2} = 6$ ;  $\frac{\partial^2 f}{\partial x \partial y} = 0$ ;  $\frac{\partial^2 f}{\partial x \partial z} = 0$ ;  $\frac{\partial^2 f}{\partial y^2} = 4$ ;  $\frac{\partial^2 f}{\partial z^2} = 2$ . Other terms are *zero*. So  $H = \begin{pmatrix} 6 & 0 & 0 \\ 0 & 4 & 0 \\ 0 & 0 & 2 \end{pmatrix}$ . Positive values, diagonal matrix. So  $f$  is positive definite — from the description  $x^T H x > 0$  for all  $x \neq 0$ . Positive **definite** and Semi **definite** matrices are very useful!

Q: For what values of  $\lambda$  is the following function convex on  $\mathbf{R}^3$ :  $f(x,y,z) = -x^2-y^2-2z^2+\lambda xy$ ? A: The Hessian is  $\begin{pmatrix} -2 & \lambda & 0 \\ \lambda & -2 & 0 \\ 0 & 0 & -4 \end{pmatrix}$  and so will not be *positive definite* for any value of  $\lambda$ . Is the function *concave* somewhere?

➤ 1st November 1999

## Section 2: Integration

### Preliminaries

We are given  $[a,b]$ . A partition  $P$  of  $[a,b]$  is a **finite** set of points  $P = \{x_0, \dots, x_n\}$ , with  $a = x_0 < x_1 < \dots < x_n = b$ . The **norm** of  $P$  is the *largest* of the numbers  $\Delta x_k = x_k - x_{k-1}$ , and is written as  $|P|$ . A partition  $P^1$  is *finer* than  $P$  (or is a refinement of  $P$ ) if  $P \subset P^1$  (so that  $P^1$  has some *extra* points). Note:  $(P \subset P^1) \Rightarrow (|P| \geq |P^1|)$ .

We **have** (we are given)  $\alpha: [a,b] \rightarrow \mathbf{R}$  which is *bounded* on  $[a,b]$ . (There exists an  $M$  such that  $|\alpha(x)| \leq M$  for all  $x \in [a,b]$ ). We write  $\Delta \alpha^k = \alpha(x_k) - \alpha(x_{k-1})$  and  $\sum_{k=1}^n \Delta \alpha_k = \alpha(b) - \alpha(a)$ .

### Definition of the Riemann-Stieltjes Sum (R-S Sum) of $f$ w.r.t. $\alpha$

Suppose that  $P$  is a *partition* of  $[a,b]$  and that  $t_k \in [x_{k-1}, x_k]$ . A sum of the form  $S(P, f, \alpha) = \sum_{k=1}^n f(t_k) \Delta \alpha_k$  is called an R-S sum *of  $f$  with respect to  $\alpha$* . We say that “ $f$  is integrable w.r.t.  $\alpha$  on  $[a,b]$ ” if there is a number  $A$  such that: for every  $\epsilon > 0$ , there is a partition  $P_\epsilon \in \mathcal{P}$   $[a,b]$ , (where  $\mathcal{P}$  is a curly  $P$ ), the set of *partitions* of  $[a,b]$ , such that if  $P$  is finer than  $P_\epsilon$ , and for every *choice* of the  $t_k$ , then  $|S(P, f, \alpha) - A| < \epsilon$ .

If such an  $A$  exists, it is unique, and we **write**  $\int_a^b f d\alpha$ . We also write  $f \in R(\alpha)$  if  $\int_a^b f d\alpha$  exists. In the case that  $\alpha(x) = x$ , this is just the *Riemann integral*,  $\int_a^b f dx$ , and we write  $f \in R$ .

**Properties:** (1) If  $f, g \in R(\alpha)$  on  $[a,b]$ , and if  $c_1, c_2 \in \mathbf{R}$ , then  $c_1 f + c_2 g \in R(\alpha)$  on  $[a,b]$  and  $\int_a^b (c_1 f + c_2 g) d\alpha = c_1 \int_a^b f d\alpha + c_2 \int_a^b g d\alpha$ . (2) If  $f \in R(\alpha)$  and  $f \in R(\beta)$ , and we have  $c_1, c_2 \in \mathbf{R}$ , then  $f \in R(c_1 \alpha + c_2 \beta)$  and (of course)  $\int_a^b f d(c_1 \alpha + c_2 \beta) = c_1 \int_a^b f d\alpha + c_2 \int_a^b f d\beta$ . (3) If  $a < c < b$ , and if **two** of the three integrals below exist, then *so does the third*, and we have  $\int_a^c f d\alpha + \int_c^b f d\alpha = \int_a^b f d\alpha$ . (4) Integration by **Parts**. If  $f \in R(\alpha)$  on  $[a,b]$ , and if  $\alpha \in R(f)$  on  $[a,b]$ , then  $\int_a^b f d\alpha + \int_a^b \alpha df = f(b)\alpha(b) - f(a)\alpha(a)$ .

**Reduction to Riemann Integrals:** Suppose that  $f \in R(\alpha)$  on  $[a,b]$ , and that  $\alpha$  has a *continuous derivative*  $\alpha'$  on  $[a,b]$ , then  $\int_a^b f(x)\alpha'(x)dx$  exists and *equals*  $\int_a^b f d\alpha$ . **Change of Variable:** Suppose that  $f \in R(\alpha)$  on  $[a,b]$ , and that  $g$  is *strictly monotone* (strictly  $\Uparrow$ ), then if  $x > x'$ , we have  $g(x) > g(x')$ . (Similarly for  $\Downarrow$  and continuous defined on  $[c,d]$ ). Now assume that  $g(c) = a$  and that  $g(d) = b$ . Let  $h(x) = f(g(x))$  and  $\beta(x) = \alpha(g(x))$  for  $x \in [c,d]$ . Then  $h \in R(\beta)$  and  $\int_c^d h d\beta = \int_a^b f d\alpha$ .

What happens with *discontinuities*? A diagram is shown of an e.g.  $\alpha$  example of a step function. Here, for  $a \leq x < c$ , we have  $\alpha(a)$ ; for  $x = c$  we have  $\alpha(c)$ , and for  $c < x \leq b$ , we have  $\alpha(b)$ . Assume we have  $f: [a,b] \rightarrow \mathbf{R}$ , and at  $c$ , *one of  $f$  and  $\alpha$*  is continuous from the left at  $c$ ; and at **least** one is continuous from the right. Then  $f \in R(\alpha)$  on  $[a,b]$ , and  $\int_a^b f d\alpha = f(c)[\alpha(c_+) - \alpha(c_-)]$  (This is the case  $f = \text{continuous}$ ).

**Idea:** Let  $P \in \mathcal{P}[a,b]$ , and assume that  $c \in P$  w.l.o.g. (w.l.o.g. = *without loss of generality*). Then,  $S(P,f,\alpha) = \sum f(t_k) \Delta \alpha_k = f(t_{k-1})[\alpha(c) - \alpha(c_-)] + f(t_k)[\alpha(c_+) - \alpha(c)]$ . (See *diagram*). Now let  $|P| \rightarrow 0$ , so that  $t_k \rightarrow c$  from the right, and  $t_{k-1} \rightarrow c$  from the left. For simplicity, **assume that  $f$  is continuous** As  $|P| \rightarrow 0$ ,  $f(t_k) \rightarrow f(c)$ , and  $f(t_{k-1}) \rightarrow f(c)$ . So  $S(P,f,\alpha) \rightarrow f(c)(\alpha(c_+) - \alpha(c_-))$ .

➤ 8th November 1999

Define a **step** function. Let  $\alpha$  be *defined* on  $[a,b]$  so that it is discontinuous at a finite number of points  $c_k$ . ( $a \leq c_1 < c_2 < \dots < c_n \leq b$ , and ' $a$ ' is a *constant* on each subinterval  $(c_{k-1}, c_k)$ ). In this case,  $\alpha$  is called a step function. The **number**  $\alpha(c_{k+}) - \alpha(c_{k-})$  is called the *jump* at  $c_k$ . If  $c_1 = a$ , the **jump** at  $c_1$  is  $\alpha(c_{1+}) - \alpha(c_1)$ . If  $c_n = b$ , the jump at  $c_n$  is  $\alpha(c_n) - \alpha(c_{n-})$ .

**Theorem:** (Reduction of an R-S integral to a *finite* sum): If  $\alpha$  is a step function on  $[a,b]$ , with jump  $\alpha_k$  at  $c_k$ , so that not *both*  $f$  and  $\alpha$  are discontinuous from the **left** or from the **right** at  $c_k$ , then  $\int_a^b f d\alpha$  exists, and equals  $\sum_{k=1}^n f(c_k) \alpha_k$ . Example: *Greatest* integer function:  $\lfloor x \rfloor = n$  if  $n \leq x$ , and  $n$  is the **largest** such  $n \in \mathbf{Z}$ . Suppose that  $\sum_{k=1}^n a_k$  is a *finite* sum.

**Define**  $f$  on  $[0, n]$  by  $f(x) = a_k$  if  $k-1 < x \leq k$ , and  $f(x) = 0$  if  $x = 0$ . Then  $\sum_{k=1}^n a_k = \sum_{k=1}^n f(k) = \int_0^n f(x) d\lfloor x \rfloor$ . **Application:** *Euler's summation formula*. If  $f$  has a *continuous* derivative  $f'$  on  $[a,b]$ , then  $\sum_{a < n \leq b} f(n) = \int_a^b f(x) dx + \int_a^b f'(x)(x - \lfloor x \rfloor) dx + f(a)(a) - f(b)(b)$ , where  $(x) = x - \lfloor x \rfloor$  is the *decimal* part function. When  $a$  and  $b$  are **integers**, this gives  $\sum_{n=1}^b f(n) = \int_a^b f(x) dx + \int_a^b f'(x)(x - \lfloor x \rfloor)^{1/2} dx + f(a) - f(b)/2$ .

**Proof:** Use *integration* by parts for  $f$ , and  $\alpha(x) = x - \lfloor x \rfloor$ . For example,  $\sum_{k=1}^n 1/k^s = (1/n^{s-1}) + s \int_1^n (\lfloor x \rfloor / x^{s+1}) dx$ . ( $s \neq 1$ ). So  $\sum_{k=1}^n 1/k = \ln(n) + C + O(1/n)$  (where  $C$  is a *constant* and  $O(1/n)$  is "of order  $1/n$ "). This is very useful in *analytic* number theory.

## Functions of Bounded Variation

Let  $P \in \mathcal{P}[a,b]$ ,  $f: [a,b] \rightarrow \mathbf{R}$ , and write  $\Delta f_k = f(x_k) - f(x_{k-1})$ . If there is a *positive number*  $M$  such that  $\sum |\Delta f_k| < M$  for all partitions of  $[a,b]$ , then  $f$  is of **bounded variation** on  $[a,b]$ . For example, any *monotone* function is of Bounded Variation (BV) on  $[a,b]$ . Example: **define**  $V_f(a,b) = \sup_{P \in \mathcal{P}[a,b]} (\sum |\Delta f_k|)$ . The total *variation* of  $f$  on  $[a,b]$  is  $V(x) = V_f(a,x)$  for  $a < x < b$ ;  $V(a) = 0$ . Therefore,  $V(x)$  is **monotonically increasing** on  $[a,b]$  and  $V - f$  is *monotonically increasing* on  $[a,b]$ . So  $f$  is of BV on  $[a,b]$  iff  $f = g - h$  for two *monotonically increasing* functions on  $[a,b]$ . (Since  $f = V - (V - f)$ ).

## Workshop (Weeks 6-7)

Finding **maxima** of functions: non-linear programming. We will look at ways of finding maxima (or minima) of *functions of many variables*. The method is known as the method of steepest ascent. It is used a lot within **neural** networks to search for stable/turning points of an “energy” function which will give a *reasonable* solution to an optimisation problem. Unless the function is concave, then this will only be a local optimum, but the method DOES usually improve guesses or estimates.

The method uses the **Gradient**: given  $f: \mathbf{R}^n \rightarrow \mathbf{R}$ , we set  $\nabla f(\underline{x}) = (\delta f / \delta x_1(\underline{x}), \dots, \delta f / \delta x_n(\underline{x}))$  to be the *gradient* vector of  $f$ . We will usually **normalise** this to get its direction,  $\nabla f(\underline{x}) / \|\nabla f(\underline{x})\|$ . Q: Find the **gradient** when  $f(x_1, x_2) = x_1^2 + x_2^2$ , and hence the *direction* of  $\nabla f(3, 4)$ . What does the *graph* of the curve  $f(x_1, x_2) = f(3, 4)$  look like? Represent the **direction** of  $\nabla f(3, 4)$  on it as a *vector* based at  $(3, 4)$ .

A:  $\delta f / \delta x_1 = 2x_1$ ,  $\delta f / \delta x_2 = 2x_2$ . So  $\nabla f(x) = (2x_1, 2x_2)$ , with  $\nabla f(3, 4) = (6, 8)$ .  $\|\nabla f(3, 4)\| = \sqrt{(6^2 + 8^2)} = \sqrt{100} = 10$ . So the **direction** is  $(\frac{6}{10}, \frac{8}{10})$ . Now  $f(x_1, x_2) = f(3, 4)$  implies that  $x_1^2 + x_2^2 = 25$ . This is a circle of *radius* 5, centred at the origin. The direction of  $\nabla f(3, 4)$  at  $(3, 4)$  will be *perpendicular* to the **tangent** to the circle at that point.

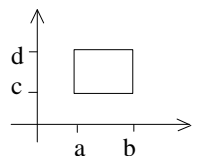
**Investigate** in general: If we move  $x$  a small length  $\delta$  in the direction defined by a unit (column) vector  $\underline{d}$  (so that  $\underline{x}$  changes to  $\underline{x} + \delta \underline{d}$ ), estimate the *change* in  $f$ , i.e. estimate  $f(\underline{x} + \delta \underline{d}) - f(\underline{x})$  in terms of  $\nabla f$ ,  $\underline{d}$  and  $\delta$ . **Show** that if  $\nabla f(\underline{x}) \cdot \underline{d} / \|\nabla f(\underline{x})\| > 0$ , then moving in the *direction* of  $\delta$  will **increase** the value of  $f(\underline{x})$ , but if it is  $< 0$ , we can expect  $f(\underline{x})$  to *decrease*.

➤ 9th November 1999

**Theorem:** (a) If  $f$  is *continuous* on  $[a, b]$ , and  $\alpha \in BV$  on  $[a, b]$ , then  $f \in R(\alpha)$  on  $[a, b]$ . (b) If  $f$  is *continuous* on  $[a, b]$ , then  $f \in R$  on  $[a, b]$ . (*Riemann* integrable). (c) If  $f$  is of *BV* on  $[a, b]$ , then  $f \in R$  on  $[a, b]$ .

## Interchanging the Order of Integration

Let  $R = \{(x, y) \mid a \leq x \leq b, c \leq y \leq d\} = [a, b] \times [c, d]$ . Assume  $\alpha$  is of *BV* on  $[a, b]$ ;  $\beta$  is of *BV* on  $[c, d]$ ; and  $f$  is *continuous* on  $R$ . Set  $F(y) = \int_a^b f(x, y) d\alpha(x)$ , and  $G(x) = \int_c^d f(x, y) d\beta(y)$ . Then  $F \in R(\beta)$  on  $[c, d]$  and  $G \in R(\alpha)$  on  $[a, b]$  imply that  $\int_a^b (\int_c^d f(x, y) d\beta(y)) d\alpha(x) = \int_c^d (\int_a^b f(x, y) d\alpha(x)) d\beta(y)$ .



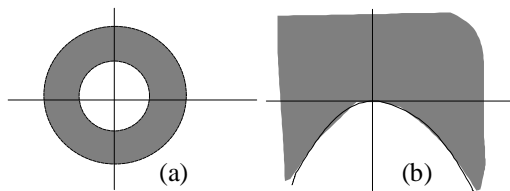
**Measure.** Let  $S \in \mathbf{R}$ . By a *Lebesgue* covering of  $S$ , we mean a collection  $T = \{I_1, I_2, \dots\}$  of open *intervals* so that  $\cup_{k=1}^{\infty} I_k \supseteq S$ . Let  $L(I_k)$  be the *length* of  $I_k$ , and set  $L(T) = \sum_{k=1}^{\infty} L(I_k)$ , provided *this* is  $< \infty$ . Set  $\bar{m}(S) = \text{g.l.b.}\{L(T) \mid T \text{ is a Lebesgue covering of } S\}$ . This is called the **outer** Lebesgue measure of  $S$ . If  $\bar{m} = 0$ ,  $S$  is said to be of *measure* zero.

**Theorem** (Lebesgue). If  $f$  is *defined* and *bounded* on  $[a,b]$ ; and let  $D$  denote the set of **discontinuities** of  $f$  in  $[a,b]$ , then  $f \in \mathbf{R}$  on  $[a,b]$  iff  $m(D) = 0$ . **Applications:** (1) If  $f \in \mathbf{R}$  on  $[a,b]$ , then  $f \in \mathbf{R}$  on  $[c,d]$  for any *subinterval*  $[c,d] \subset [a,b]$ ,  $|f|$  is  $\mathbf{R}$  on  $[a,b]$ , and  $f^2 \in \mathbf{R}$  on  $[a,b]$ ,... (2) If  $f \in \mathbf{R}$  and  $g \in \mathbf{R}$  on  $[a,b]$ , then *provided*  $g$  is bounded away from 0 on  $[a,b]$ , then  $f/g \in \mathbf{R}$  on  $[a,b]$ .

## Assignment 1

Q: For (a)  $f(x,y) = \ln\{(25-x^2-y^2)(x^2+y^2-4)\}$  and (b)  $f(x,y) = \sqrt{x^2+2y}$ , give (i) the *domain*, (ii) sketch  $D_f$  in the usual  $x$ - $y$  plane, (iii) give the *image* or *range* of  $f$ : A: (a) (i) The things in the brackets must be *positive*, so  $25 > x^2+y^2 > 4$  **or**  $25 < x^2+y^2$  and  $x^2+y^2 < 4$  (oops: impossible!). So we need the first one. Such points satisfying this form an *annulus* with outer radius 5; inner radius 2.

(iii) For any  $(x,y) \in D_f$ ,  $(25-(x^2+y^2))(x^2+y^2-4) > 0$ . But setting  $r^2 = x^2+y^2$ , what is the maximum value of the expression  $(25-r^2)(r^2-4) = 29r^2-r^4-100 = F(r)$ , say?  $dF/dr = 58r-4r^3 = 0$  for maximum or minimum. This has *two* solutions:  $r = 0$  or  $r^2 = 58/4 = 14.5$ , i.e.  $r \approx 3.8$ . Thus  $0 < F(r) \leq (25-14.5)(14.5-4) = (10.5)^2 = 110.25$ . Thus  $-\infty < f(x,y) \leq \ln(110.25)$ . (Since  $F(r) > 0$  but we can get *arbitrarily* near to 0).



(b) We need  $x^2+2y \geq 0$ , i.e.  $y \geq -1/2x^2$ . So  $D_f = \{(x,y) \mid y \geq -1/2x^2\}$ .  $A = \sqrt{x^2+2y}$  can take any *non-negative* value, (even keeping  $y = 0$  we get  $\sqrt{x^2}$ !), so the **image** of  $f$  is  $\mathbf{R}^+$ . Note that when talking about what values the *domain* has, use “**accepts**” rather than “**takes**” and “consists of” rather than “contains”. Also use  $[a,b]$  etc., for *specifying* ranges, not  $-5 < x < 6$ , etc.

Q: Does the **function**  $f(x,y) = (x^2+y)/(3x^4+2y^2)$  have a limit at the *origin*? Justify your answer. (ii) Show that  $\lim_{x \rightarrow 1, y \rightarrow 2} (x^2+y^2) = 5$ . A: Suppose one looks *along* the  $y$ -axis, (so  $x = 0$  here), then  $f(0,y) = 1/2y$ . This tends to  $\pm\infty$  as  $y$  tends to 0, so  $f$  **cannot** have a limit at the origin.

(ii) (Plan ahead — we want to end up with a statement: “*given*  $\epsilon > 0$ , there is a  $\delta$  such that if  $\|(x,y)-(1,2)\| < \delta$ , then  $|f(x,y)-5| < \epsilon$ ”. If  $\|(x,y)-(1,2)\| < \delta$ , then  $|x-1| < \delta$  and  $|y-2| < \delta$ . See what this implies for a *general*  $\delta$ . If  $|x-1| < \delta$ , then  $1-\delta < x < 1+\delta$ , and we expect  $\delta < 1$ , so in this *interval*,  $x$  is positive. Thus  $1-2\delta < 1-2\delta+\delta^2 < x^2 < 1+2\delta+\delta^2 < 1+3\delta$ .

*Similarly*, if  $|y-2| < \delta$ , then  $4-4\delta < y^2 < 4+5\delta$ . Putting these together gives  $|x^2+y^2-5| < 8\delta$ , and the only *restriction* was  $\delta < 1$ . To get this  $< \epsilon$ , choose  $\delta \leq \epsilon/10$ . Now put the building blocks together. Here is a *possible* way: Given  $\epsilon > 0$ , set  $\delta = \min(\epsilon/10, 1)$ . If  $\|(x,y)-(1,2)\| < \delta$ , then (a)  $1-\delta < x < 1+\delta$ , and (b)  $2-\delta < y < 2+\delta$ . (a) implies that  $-2\delta < x^2-1 < 3\delta$  whilst (b) implies that  $4\delta < y^2-4 < 5\delta$ . Hence  $|x^2+y^2-5| < 8\delta < \epsilon$  as required, i.e.  $\lim_{x \rightarrow 1, y \rightarrow 2} (x^2+y^2) = 5$ .

Q: Suppose that  $X$  is a **convex** set in  $\mathbf{R}$  (so is an interval). Suppose that  $f: X \rightarrow \mathbf{R}^1$  is a convex function, and that  $x_0$  is a point at which  $f$  has a local *minimum*. (No assumption about the differentiability of  $f$  is needed: “ $x_0$  is a local minimum” means “there is some  $\varepsilon > 0$  such that if  $x \in X$  and  $|x-x_0| < \varepsilon$ , then  $f(x_0) \leq f(x)$ ”). Prove that  $f(x_0)$  is a *global minimum* (so no value of  $f$  is less than this one).

A: Suppose that  $X$  is **convex**;  $X \subseteq \mathbf{R}$ ;  $f: X \rightarrow \mathbf{R}^1$  is a *convex* function; and  $x_0$  is a point at which  $f$  has a local **minimum**. “ $X$  is convex” means “if  $x, x' \in X$  are any *two* points, and  $t \in [0,1]$ , then  $x_t = tx+(1-t)x' \in X$ ”. “ $f$  is a *convex* function” means “if  $x, x' \in X$ , and  $t \in [0,1]$ , then  $f(x_t) \leq tf(x)+(1-t)f(x')$ ”. “ $f$  has a *local minimum* at  $x_0$ ” means “if  $x, x' \in X$  and  $t \in [0,1]$  such that  $|x-x_0| < \varepsilon$ , then  $f(x_0) \leq f(x)$ ”.

The idea of the proof is to go from an arbitrary  $x$  along the line to  $x_0$ , until we get to near  $x_0$  — where we know about the **relative** sizes of  $f(x_t)$  and  $f(x_0)$  by the above. Then we use *algebra*. **Proof:** Suppose that  $x \in X$  ( $x \neq x_0$ ). We want to show that  $f(x) \geq f(x_0)$ . As  $x \in X$ , we know that for any  $t \in [0,1]$ ,  $x_t = tx+(1-t)x_0$  is also in  $X$ .

If we pick  $0 < t < \varepsilon/|x-x_0|$ , where  $\varepsilon$  is given by the *local* maximality condition (above), then  $|x_t-x_0| = t(x-x_0) < \varepsilon$ . So by the above,  $f(x_t) \geq f(x)$  (---(a)). By the *convexity* of  $f$ ,  $tf(x)+(1-t)f(x_0) \geq f(x_t)$ . *Rearranging* this gives  $tf(x) \geq tf(x_0)$ . But we (cunningly) picked  $t > 0$ , so this *implies* that  $f(x) \geq f(x_0)$  as required.

Q: Determine whether each of the *following* functions is convex, concave or neither. Then find and classify any **stationary** points. (i)  $f(x_1, x_2) = 2x_1^3 + 3x_1x_2 + x_2^2$ ,  $x \in \mathbf{R}^2$ ; (ii)  $f(x_1, x_2, x_3) = x_1^2 + x_2^2 - 2x_3^2 + 0.5x_1x_2$ ,  $x \in \mathbf{R}^3$ . A:  $f$  is convex iff for all  $x \in S$ , all *principle minors* are  $\geq 0$ .  $f$  is concave iff all non-zero principal  $k \times k$  minors have the same sign as  $(-1)^k$ .

(i)  $H = \begin{pmatrix} 12x_1 & 3 \\ 3 & 2 \end{pmatrix}$ . So as  $x_1$  could be +ve or -ve, this is not *going* to be “definite” at all. More exactly, if  $12x_1 \geq 0$ , we might hope for +ve definite, but  $\det(H) = 24x_1 - 9$ , so  $f$  is only convex on the set  $x_1 \geq 9/24$ . Stationary Points:  $f$  has a local min at  $(3/4, -9/8)$ , and a saddle point at  $(0,0)$ . (ii)  $H = \begin{pmatrix} 2 & 0.5 & 0 \\ 0.5 & 2 & 0 \\ 0 & 0 & -4 \end{pmatrix}$ . The 1st principle minors are 2, 2 and -4, so the function is neither convex **nor** concave.

Q: For what *values* of  $a$ ,  $b$  and  $c$  is  $ax_1^2 + bx_1x_2 + cx_2^2$  a *convex* function on  $\mathbf{R}^2$ ? A:  $H = \begin{pmatrix} 2a & b \\ b & 2c \end{pmatrix}$ . We need  $a \geq 0$ ;  $c \geq 0$  and  $b^2 - 4ac \leq 0$ .

Q: Find all *local minima*, *local maxima* and *saddle* points for  $f(x_1, x_2) = x_1^3 - x_2^2 + x_1x_2 - 3x_2$  on  $\mathbf{R}^2$ . A:  $\delta f / \delta x_1 = 3x_1^2 + x_2$ . If  $\delta f / \delta x_1 = 0$ , then  $x_2 = -3x_1^2$  (---(\*)).  $\delta f / \delta x_2 = -2x_2 + x_1 - 3$ . If  $\delta f / \delta x_2 = 0$ , then  $x_2 = 1/2(x_1 - 3)$  (---(\*\*)). *Solving* (\*) & (\*\*), we get  $6x_1^2 + x_1 - 3 = 0$ ;  $x_1 = \frac{-1 \pm \sqrt{1+72}}{12} = \frac{-1 \pm \sqrt{73}}{12}$ . We will need *approximate values* of this:  $x_1 \approx -0.795$  or  $0.629$ .

Let us write  $\underline{x}^{(1)}$  and  $\underline{x}^{(2)}$  for  $\underline{x}^{(1)} = (-1-\sqrt{73}/12, -37/24-\sqrt{73}/24)$  and  $\underline{x}^{(2)} = (-1+\sqrt{73}/12, -37/24+\sqrt{73}/24)$ . The *Hessian*,  $H(\underline{x})$ , is  $H(\underline{x}) = \begin{pmatrix} 6x_1 & 1 \\ 1 & -2 \end{pmatrix}$ . The *principle minors* are as follows: (1st):  $6x_1$  and  $-1$ ; (2nd):  $-12x_1 - 1$  (i.e.  $\det(H(\underline{x}))$ ).

At  $\underline{x}^{(1)}$ , both the first principal *minors* are negative (since  $x_1^{(1)} \approx -0.8$ ), whilst the *second* principal minor is approximately 6.2. The **signs** are also  $(-1)^k$ : the 1st ones ( $k = 1$ ) are *negative*; the 2nd one ( $k = 2$ ) is *positive*. Thus  $H(\underline{x})$  is positive definite **near**  $\underline{x}^{(1)}$ , and  $f$  has a *local* maximum at  $\underline{x}^{(1)}$  ( $f$  is *concave* at  $\underline{x}^{(1)}$ ). At  $\underline{x}^{(2)}$ ,  $6x_1^{(2)}$  is positive ( $\approx 3.6$ ), but  $-2$  is *negative*. The 2nd principal minor is negative. Thus it is a **saddle** point.  $f$  is neither concave *nor* convex near  $\underline{x}^{(2)}$ . Alternatively, apply the earlier test. This is really the same here as we have 2 variables.

➤ 15th November 1999

## Workshop (Weeks 6 & 7) Continued

**Method of steepest ascent.** Begin at some point  $\underline{v}_0$ . For non-negative  $t$ , let  $\underline{v}_t = \underline{v}_0 + t\nabla f(\underline{v}_0)$ . (i) Find  $t_0 \geq 0$  such that  $t_0$  *maximises*  $f(\underline{v}_t)$ . Set  $\underline{v}_1 = \underline{v}_{t_0}$ . (ii) Ask “Is  $\|\nabla f(\underline{v}_1)\|$  *smaller* than some preset bound?”, e.g. 0.01. If “yes”, **terminate**, as  $\nabla f(\underline{v}_1)$  being that small means that  $\underline{v}_1$  is *approximately* a stationary point. If “no”, then move from  $\underline{v}_1$  in the direction of  $\nabla f(\underline{v}_1)$ , and repeat. When the process *stops*, with  $\underline{v}_n$ , say, then  $\underline{v}_n$  is an *approximate* stationary point of  $f$ .

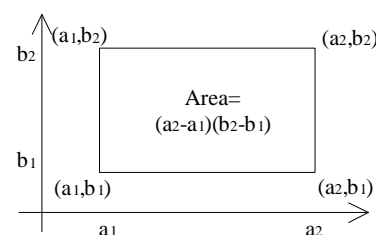
**Predicting change with differentials.** Suppose we know that  $f: S \rightarrow \mathbf{R}$  is differentiable, and also know  $f(\underline{x}_0)$  and  $Df(\underline{x}_0)$ . Can we estimate  $f(\underline{x}_0 + \Delta \underline{x})$  for *small* values of  $\Delta \underline{x}$ ? If  $\Delta \underline{x}_0$  is small, (i.e.  $\|\Delta \underline{x}_0\|$  is small), then  $f$  and its *linear approximation* will not differ by **very** much, and we can estimate the *change*  $\Delta f = f(\underline{x}_0 + \Delta \underline{x}) - f(\underline{x}_0)$  to be  $L(\underline{x}_0 + \Delta \underline{x}) - L(\underline{x}_0)$ , where  $L(\underline{x}) = f(\underline{x}_0) + Df(\underline{x}_0)(\underline{x} - \underline{x}_0)$ .

Of course, this gives an *estimate* for  $\Delta f$  given by  $\nabla f(\underline{x}_0) \cdot \Delta \underline{x}$ . As this is a good *estimate* “in the limit as  $\|\Delta \underline{x}_0\| \rightarrow 0$ ”, we use *differential* notation:  $df = \nabla f(\underline{x}_0) \cdot d\underline{x}$ , where  $d\underline{x} = (dx_1, \dots, dx_n)$ , if a *co-ordinate* system is given. In other words,  $df = \sum_{i=1}^n (\delta f / \delta x_i) dx_i$ . Q: Let  $z = (x/y)^{1/3}$ . Find the approximate **maximum** percentage error in measuring  $z$  if there is a *possible* 1% error in measuring  $x$ , and a *possible* 0.5% error in measuring  $y$ .

➤ 15th November 1999

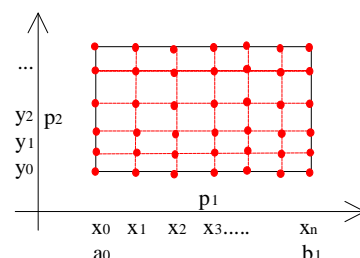
## Multiple Integrals

To **generalise** integration to many variables, we will need a *replacement* for the length of an interval  $[a, b]$  (i.e.  $(b-a)$ ) for many dimensions. In dimension 3, we use the area of a **rectangle** as shown. In 3 dimensions, the volume is  $(a_2 - a_1)(b_2 - b_1)(c_2 - c_1)$ . What is the measure “A” of a *bounded*  $n$ -dimensional interval in  $\mathbf{R}^n$ , if  $A_1, \dots, A_n$  are *bounded* intervals in  $\mathbf{R}$ ?



A:  $A = A_1 \times \dots \times A_n = \{(x_1, \dots, x_n) \mid x_k \in A_k, k = 1, \dots, n\}$ . The  $A_i$  may be *open*, *closed* or *half-open* (but will usually be **bounded**). We say that  $A$  is an  *$n$ -dimensional interval*. Each  $A_k$  has *finite* length, e.g.  $A_k = [a_k, b_k]$ , where  $b_k - a_k < \infty$ . We write  $\mu(A_k) = b_k - a_k$ . Let us *define*  $\mu(A) = \mu(A_1) \dots \mu(A_n)$ , the *measure* of  $A$ .

**Partitions of A.** Let  $A = A_1 \times \dots \times A_n$  be closed and bounded. If  $P_k$  is a partition of  $A_k$ , let  $P = P_1 \times \dots \times P_n$ . ( $P$  is a *partition* of  $A$ ). It is **fairly** clear **when**  $P' < P$  (“ $<$ ” = finer than) occurs, because we are using *partitions* of  $A$ . ( $P \subseteq P'$ ).



**Riemann Sum:** Suppose that  $f$  is *defined* and *bounded* on an  $I \subset \mathbf{R}$ . ( $I$  is a *generalised* closed bounded subinterval of  $\mathbf{R}^n$ ). If  $P \in \mathcal{P}(I)$ , ( $\mathcal{P}(I) =$  partitions of  $I, \dots, I_m$ ), where  $P$  divides  $I$  into  $m$  subintervals  $I_k$ ; and  $t_k \in I_k$ , then  $S(P, f) = \sum_{k=1}^m f(t_k)\mu(I_k)$  is called a *Riemann sum*.

## Riemann Integrable Functions

$f$  is said to be *Riemann integrable* on  $I$  ( $f \in R$  on  $I$ ) whenever there is a *real number*  $M$  such that for every  $\epsilon > 0$ , then **there** is a  $P_\epsilon \in \mathcal{P}(I)$  such that if  $P < P_\epsilon$ , and  $|S(P, f) - M| < \epsilon$  for all *Riemann sums*  $S(P, f)$ , then  $M$  is *denoted by*  $\int_I f d\mathbf{x}$ , or  $\int_I f(\mathbf{x})d\mathbf{x}$ , or  $\int_I f(x_1, \dots, x_n)dx_1 dx_2 \dots dx_n$ . Also, for **small**  $n$ , it can be written as a “*multiple integral*”, e.g.  $\int_I f(x_1, x_2)dx_1 dx_2$ .

## Multiple Integrals: Theory

The **basic** idea is to first integrate w.r.t. the first variable  $dx_1$ , pretending all other variables are constants, so that you get an *expression* without  $x_1$ 's in it. Then you repeat with  $x_2$ , etc. This is relatively easy when the region you are integrating is a **rectangle**. But often, the limits of integration of the *inner* integral depend on another variable. Luckily, this does not make much difference — when you come to substitute into the **expression** for the integral, you just substitute those “*variable values*”.

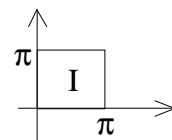
Often, the integral is not in a form which is **easy** to evaluate. Write the region in another form if the problem is the order of integration. For example, if the *inner integral* is w.r.t.  $y$ , with the region expressed with a constraint on  $y$  in terms of  $x$ , draw the region, and rework the constraint to give a constraint on  $x$  in terms of  $y$ . Now try the integral the “*other way around*”, integrating first w.r.t.  $x$  and then w.r.t.  $y$ . This **generalises** to higher dimensions, but is harder — you cannot draw *higher* dimensional regions!

➤ 16th November 1999

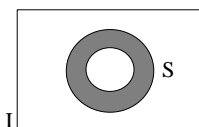
A **subset**  $T \subset \mathbf{R}^n$  is said to be of *measure zero* if for every  $\epsilon > 0$ ,  $T$  can be *covered* by a collection  $\{I_k\}_{k=1}^\infty$  of *n-dimensional intervals*, with  $\sum \mu(T_k) < \epsilon$ . **Theorem:** If  $f$  is defined and bounded on a *closed bounded*  $I \subset \mathbf{R}^n$ , then  $f \in R$  on  $I$  iff the set of *discontinuities* of  $f$  has measure zero.

## Example of Multiple Integrals by Iterated Integrations

Example:  $I = [0, \pi] \times [0, \pi]$ ,  $f(x, y) = \sin^2 x \sin^2 y$ . So  $\int_I \sin^2 x \sin^2 y dy dx = \int_0^\pi (\int_0^\pi \sin^2 x \sin^2 y dy) dx = \int_0^\pi \sin^2 x (\int_0^\pi \sin^2 y dy) dx = \int_0^\pi \sin^2 x [\int_0^\pi (1 - \cos 2y)/2 dy] dx = \int_0^\pi \sin^2 x [y/2 - \sin 2y/4]_0^\pi dx = \pi/2 \int_0^\pi \sin^2 x dx = (\pi/2)^2$ .



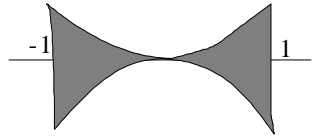
Could we have **handled** an integral over some *less regular region* — other than a rectangle? For example, if  $S = \{(x, y) \mid 1 \leq x^2 + y^2 \leq 2\}$  and  $f(x, y) = \sqrt{x^2 + y^2 - 1}$ , what should  $\int_S f dx dy$  mean? Idea: Pick some  $I \supset S$  (in *general*) and let  $\chi_S: I \rightarrow \mathbf{R}$  be defined by  $\chi_S(s) = \{1 \text{ if } s \in S; 0 \text{ if } s \notin S\}$ , the *characteristic function* of  $S$ . If  $\chi_S$  is *Riemann integrable* on  $I$ , then set  $g: I \rightarrow \mathbf{R}$  to be  $g(\mathbf{x}) = \{f(\mathbf{x}) \text{ if } \mathbf{x} \in S; 0 \text{ if } \mathbf{x} \notin S\}$ , and define  $\int_S f d\mathbf{x} = \int_I g d\mathbf{x}$ . In our circumstances, the subset  $S$  will *always* be such that  $\int_I \chi_S dx$  exists, but more *general cases* can arise.



# Order of Integration

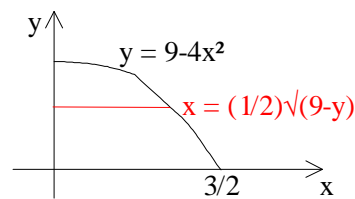
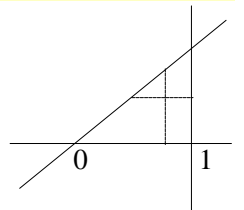
**Fubini's Theorem:** (but with a *rectangular* region): If  $f(x,y)$  is *continuous* on  $I = [a,b] \times [c,d]$ , then  $\int_S f d\underline{x} = \int_c^d (\int_a^b f(x,y) dx) dy = \int_a^b (\int_c^d f(x,y) dy) dx$ . Not **all** regions are rectangular. Suppose that  $g_1, g_2: [a,b] \rightarrow \mathbf{R}$  are *continuous*, and let  $S = \{(x,y) \mid x \in [a,b], g_1(x) \leq y \leq g_2(x)\}$  (so here we are *assuming* that  $g_1(x) \leq g_2(x)$  for all  $x$ ).

**Example:** Let  $g_1(x) = -x^2$  and  $g_2(x) = x^2$  for  $x \in [-1,1]$ . Assuming that  $f: S \rightarrow \mathbf{R}$  is continuous, then  $\int_S f d\underline{x} = \int_{-1}^1 (\int_{-x^2}^{x^2} f(x,y) dy) dx$ . If  $S$  is as above, and  $f(x,y) = xe^{-y}$ , then the integral is  $\int_S xe^{-y} d\underline{x} = \int_{-1}^1 (\int_{-x^2}^{x^2} xe^{-y} dy) dx$ . **Evaluate** the inner bracket:  $\int_{-x^2}^{x^2} xe^{-y} dy = x \int_{-x^2}^{x^2} e^{-y} dy = x[-e^{-y}]_{-x^2}^{x^2} = x(-e^{x^2} + e^{-x^2})$ . So the integral is  $\int_{-1}^1 x(e^{-x^2} - e^{x^2}) dx = \dots$  etc.



➤ 23rd November 1999

Here, the region worked well with the *order* of integration. This does not **always** happen. Example: Let  $S$  be the region in  $\mathbf{R}^2$  bounded by the  $x$ -axis, the line  $y = x$ , and the line  $x = 1$  (as shown). Then  $S = \{(x,y) \mid x \in [0,1], 0 \leq y \leq x\}$ . This is the same as  $S = \{(x,y) \mid y \in [0,1], y \leq x \leq 1\}$ . Take  $f(x,y) = \frac{\sin(x)}{x}$ . Look at  $\int_S f d\underline{x} = \int_0^1 (\int_y^1 \frac{\sin(x)}{x} dx) dy$ . Reverse, giving  $\int_0^1 (\int_0^x \frac{\sin(x)}{x} dy) dx = \int_0^1 [\frac{\sin(x)}{x} y]_0^x dx = \int_0^1 \sin(x) dx = -\cos(1) + 1$ .



**Typical Question:** Sketch the **region** of integration and write an equivalent double integral with the order of integration *reversed*. Let  $\text{Int} = \int_0^{3/2} \int_0^{9-4x^2} 16xy dy dx = \int_0^{3/2} [8xy^2]_0^{9-4x^2} dx = \int_0^{3/2} 8x(9-4x^2)^2 dx = \dots$  **Reversing** the order, we get  $\text{Int} = \int_0^9 \int_{\frac{1}{2}\sqrt{9-y}}^1 16x dx dy = \int_0^9 [8x^2]_{\frac{1}{2}\sqrt{9-y}}^1 dy = \dots$

## Tutorial

**Q:** Evaluate the **following** integrals: (a)  $\int_{-1}^1 \int_0^2 (2x^2 + 2xy + 3y^2) dy dx$ , (b)  $\int_0^3 \int_1^2 xye^{(2x+3y)} dx dy$ .  
**A:** (a) Evaluating the *inner* integral:  $\int_0^2 2x^2 + 2xy + 3y^2 dy = [2x^2y + xy^2 + y^3]_0^2 = [4x^2 + 4x + 8]$ . Now we have  $\int_{-1}^1 4x^2 + 4x + 8 dx = [\frac{4}{3}x^3 + 2x^2 + 8x]_{-1}^1 = [(\frac{4}{3} + 2 + 8) - (-\frac{4}{3} + 2 - 8)] = \frac{34}{3} + \frac{22}{3} = \frac{56}{3}$ .

(b)  $\int_0^3 \int_1^2 xye^{(2x+3y)} dx dy = \int_0^3 \int_1^2 x e^{-2x} y e^{3y} dx dy = \int_0^3 y e^{3y} dy \int_1^2 x e^{-2x} dx$ . *Evaluate* the first integral by parts: Let  $u = y$ ,  $\frac{du}{dy} = 1$ ;  $\frac{dv}{dy} = e^{3y}$ ,  $v = \frac{1}{3}e^{3y}$ . So  $\int_0^3 y e^{3y} dy = [\frac{y}{3}e^{3y}]_0^3 - \int_0^3 \frac{1}{3}e^{3y} dy = [\frac{y}{3}e^{3y}]_0^3 - [\frac{1}{9}e^{3y}]_0^3 = [\frac{y}{3}e^{3y}]_0^3 - [\frac{1}{9}e^9 - \frac{1}{9}] = [e^9 - 0] - [\frac{1}{9}e^9 - \frac{1}{9}] = e^9 - \frac{1}{9}e^9 + \frac{1}{9} = (\frac{8}{9}e^9 + \frac{1}{9})$ . Similarly,  $\int_1^2 x e^{-2x} dx = [\frac{x}{-2}e^{-2x}]_1^2 - \int_1^2 \frac{1}{-2}e^{-2x} dx = [-\frac{x}{2}e^{-2x}]_1^2 + [\frac{1}{4}e^{-2x}]_1^2 = [-\frac{2}{2}e^{-4} + \frac{1}{4}e^{-4}] + [\frac{1}{4}e^{-2} - \frac{1}{4}e^{-2}] = [-e^{-4} + \frac{1}{4}e^{-4}] + [\frac{1}{4}e^{-2} - \frac{1}{4}e^{-2}] = (-\frac{3}{4}e^{-4} + \frac{1}{4}e^{-2})$ . So the *answer* is  $(\frac{8}{9}e^9 + \frac{1}{9})(-\frac{3}{4}e^{-4} + \frac{1}{4}e^{-2})$ .

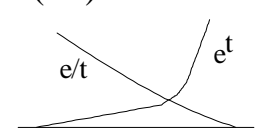
## Assignment 2

**Q:** Given the **mapping**  $x = u^2 + v^2$ ;  $y = 2uv$ , and the function  $w = w(x,y)$ , express  $\nabla^2 w$  in terms of *partial* derivatives with respect to  $u$  and  $v$ . **A:** The answer is too *detailed* to give in full here. Technique: find  $\frac{\partial u}{\partial x}$  and  $\frac{\partial v}{\partial x}$  by *implicit* differentiation; then find  $\frac{\partial^2 w}{\partial x^2}$  and  $\frac{\partial^2 w}{\partial y^2}$ . **Tips:** Keep any *factors* outside, group *related* terms, *split* the calculation into bits, and *reassemble* at the end. Look out for any *shortcuts*, e.g.  $v^2 - u^2 = -(u^2 - v^2)$ .

Q: Compute the **Jacobian** matrix of each of the following transformations, and determine where *local* inverses exist: (a)  $x = e^u \cos v$ ,  $y = \sin v$ ; (b)  $x = u^2 + v^2$ ,  $y = u + v$ . A: (a)  $J = \begin{bmatrix} e^u \cos v & -e^u \sin v \\ 0 & \cos v \end{bmatrix}$ .  $\text{Det}(J) = e^u \cos^2 v > 0$ , except when  $v = 2k+1/2\pi$ ,  $k = 0, \pm 1, \dots$ . Thus *local* inverses exist except at **odd** multiples of  $\pi/2$  in the  $v$ -direction. (b)  $J = \begin{pmatrix} 2u & 2v \\ 1 & 1 \end{pmatrix}$ .  $\text{Det}(J) = 2(u-v)$ . Local inverses exist **except** on the line  $u = v$ .

Q: Perform 2 iterations of the method of **steepest** ascent in an attempt to maximise  $f(x_1, x_2) = (x_1 + x_2)e^{-(x_1 + x_2)} - x_1$ , starting at the *point*  $(0, 1)$ . A:  $\nabla f = ((1 - (x_1 + x_2))e^{-(x_1 + x_2)} - 1, (1 - (x_1 + x_2))e^{-(x_1 + x_2)})$ .  $\underline{v}_0 = (0, 1)$ ,  $\nabla f(0, 1) = (-1, 0)$ . Note:  $\|\nabla f(\underline{v}_0)\| = 1$ . And  $\underline{v}_t^{(0)} = \underline{v}_0 + t\nabla f(\underline{v}_0) = (-t, 1)$ .  $f(\underline{v}^{(0)}) = (1-t)e^{-(1-t)} + t$ .

Next, find the **maximum** value of  $f(\underline{v}_0^{(t)})$  as a function of  $t$ . A:  $\frac{df}{dt}(\underline{v}_t^{(0)}) = (1-t)e^{-(1-t)} - e^{-(1-t)} + 1 = -te^{-(1-t)} + 1$ . For stationary values, this will be *zero*, so at  $e^t = e^{-t}$ , we have the *diagram* shown. The two curves meet at 1 point **only**, and observation says that  $t = 1$  is the answer. (The *double* derivative test shows that this is a max.).



$\underline{v}_1 = (-1, 1) = \underline{v}_1^{(0)}$ .  $\underline{v}_t^{(1)} = (-1, 1+t)$ ;  $\frac{df}{dt}(\underline{v}_t^{(1)}) = (1-t)e^{-t}$ ;  $\underline{v}_2 = (-1, 2)$ .  $\nabla f(\underline{v}_1) = (0, 1)$ ;  $f(\underline{v}_1^{(1)}) = te^{-t} + 1$ , which equals zero *when*  $t = 1$ . (Again a maximum). Note that  $\nabla f(\underline{v}_2) = (-1, 0)$ . Although there is **no** maximum, our iteration *is* increasing the value of  $f$  (see the table).

$\underline{v}$	$(0, 1)$	$(-1, 1)$	$(-1, 2)$	$(-2, 2)$
$f(\underline{v})$	$e^{-1}$	1	$1 + e^{-1}$	2

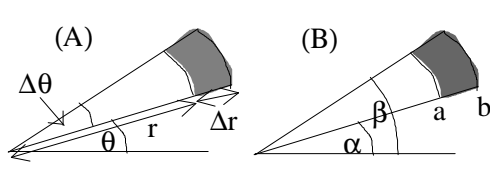
Q: **Euler's** theorem on homogenous functions is: Let  $f$  be defined on an *open* set  $S$  in  $\mathbf{R}^n$ . We say that  $f$  is *homogeneous* of degree  $p$  over  $S$  if  $f(\lambda \underline{x}) = \lambda^p f(\underline{x})$  for every real  $\lambda$  and *every*  $\underline{x} \in S$  such that  $\lambda \underline{x} \in S$ . Euler's *theorem* on homogeneous functions in part states that in this case, if  $f$  is differentiable at  $\underline{x}$ , then  $\underline{x} \cdot \nabla f(\underline{x}) = pf(\underline{x})$ . **Prove** This.

A: Let  $f$  be **homogenous** of degree  $p$  on  $S$ . Fix  $\underline{x} \in S$ . Let  $g(\lambda) = f(\lambda \underline{x})$ . By *first principles*,  $g'(1) = \lim_{h \rightarrow 0} \frac{g(1+h) - g(1)}{h} = \lim_{h \rightarrow 0} \frac{f((1+h)\underline{x}) - f(\underline{x})}{h}$  (---(\*)) = (since  $f$  is *homogenous of degree*  $p$ ) =  $\lim_{h \rightarrow 0} \frac{((1+h)^p)^{1/p} f(\underline{x})}{h} = (\dots)$  as in  $(\underline{x} + h)^p = \underline{x}^p + p\underline{x}^{p-1}h + \dots$ , the *binomial theorem* is used in the derivation of  $\frac{d}{d\underline{x}}(\underline{x}^p) = p\underline{x}^{p-1} = \lim_{h \rightarrow 0} (p + \text{terms in } h)f(\underline{x}) = pf(\underline{x})$ .

Going **back** to (---(\*)), this is  $\lim_{h \rightarrow 0} \frac{f(\underline{x} + h\underline{x}) - f(\underline{x})}{h}$ . But this is the **definition** of the directional derivative of  $f$  at  $\underline{x}$  in the direction of  $\underline{x}$ , and we have that this is  $\underline{x} \cdot \nabla f(\underline{x})$ . Putting this **together**,  $\underline{x} \cdot \nabla f(\underline{x}) = pf(\underline{x})$ , i.e. if  $\underline{x} \in \mathbf{R}^n$ , then  $\sum_{i=1}^n x_i (\partial f / \partial x_i) = pf(\underline{x})$ .

## Double Integration in Polar Co-ordinates

Looking at the **diagram** on the right, (A), the area of a segment with angle  $\Delta\theta$  and radius  $r$  is  $\frac{1}{2}\Delta\theta r^2$ . The area of  $\epsilon$  segment with angle  $\Delta\theta$  and radius  $(r + \Delta r)$  is  $\frac{1}{2}\Delta\theta (r + \Delta r)^2$ . The *shaded* area is  $r\Delta r\theta + \text{terms in } \Delta r^2\Delta\theta$ . In the *Riemann sum*



with a partition of some sector as shown, (B),  $S(P, f) = \sum f(\underline{t}_k) \mu(S_k) = \sum f(\underline{t}_k) r_{k-1} \Delta r_k \Delta \theta_k + \text{small terms}$ . As the *norm* of the partition tends to zero, this approaches  $\int_a^b \int_\alpha^\beta f(r, \theta) r dr d\theta$ .

**Example:** Area of the sector:  $\int_{\alpha}^{\beta} \int_{r_1}^{r_2} r dr d\theta = \int_{\alpha}^{\beta} \frac{1}{2}(r_2^2 - r_1^2) d\theta = \frac{1}{2}(r_2^2 - r_1^2)(\beta - \alpha)$ . We can adopt this to the *general* case of a change in co-ordinates. The basic element of *volume/area/measure* is  $dx_1, \dots, dx_n = \mathbf{determinant}$  of a matrix with  $dx_i$  on the diagonals and *zeroes* everywhere else.  $dx_i$  is an "*infinitesimal*" change in the  $x_i$  direction.

If we **change** from  $x_1, \dots, x_n$  co-ordinates to  $y_1, \dots, y_n$  co-ordinates, where  $x_k = g_k(y_1, \dots, y_n)$ , then  $dx_k = \sum_{i=1}^n \left( \frac{\partial x_k}{\partial y_i} dy_i \right) = Dg \cdot \begin{pmatrix} dy_1 \\ \vdots \\ dy_n \end{pmatrix}$ . So,  $dx_1 \dots dx_n = \det(Dg) dy_1 \dots dy_n$ . **Examples:** (1) **Polars:**  $x_1 = r \cos \theta$ ,  $x_2 = r \sin \theta$ ,  $J = \partial(x_1, x_2) / \partial(r, \theta) = r$ . The integral becomes  $\int \int f(r, \theta) r dr d\theta$ .

(2) **Cylindrical polars:**  $(x_1, x_2, x_3) \leftrightarrow (r, \theta, z)$ .  $J = r$  again.  $\int \int \int f(r, \theta, z) r dr d\theta dz$ .  
 (3) **Spherical polars:**  $(x_1, x_2, x_3) \leftrightarrow (\rho, \phi, \theta)$ .  $J = |\rho^2 \sin \phi|$  (the volume is +ve so we do need the |...|). The *integral* becomes  $\int \int \int f(\rho, \phi, \theta) \rho^2 \sin \phi d\rho d\phi d\theta$ . Look in *Thomas & Finney* for other examples.

**Example:**  $I = \int_{\Omega} xy dx dy$ . Convert to *polar* co-ordinates:  $\Omega = \{(r, \theta) \mid 0 \leq r \leq 1, 0 \leq \theta \leq \pi/2\}$ . So  $I = \int_{\theta=0}^{\pi/2} \int_{r=0}^1 r^3 \cos \theta \sin \theta dr d\theta = 1/8$ . Example: to evaluate  $\int_{-\infty}^{\infty} e^{-x^2} dx = I$ , evaluate  $I^2 = (\int_{-\infty}^{\infty} e^{-x^2} dx)(\int_{-\infty}^{\infty} e^{-y^2} dy) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{-(x^2+y^2)} dx dy$ . Let  $D_b = \{(x, y) \mid 0 \leq x^2 + y^2 \leq b^2\}$ . So  $\int_{D_b} e^{-(x^2+y^2)} dx dy = \int_{\theta=0}^{2\pi} \int_{r=0}^b e^{-r^2} r dr d\theta = (\int_0^{2\pi} d\theta) [-1/2 e^{-r^2}]_0^b = 2\pi(1 - e^{-b^2/2}) = \pi - \pi e^{-b^2}$ . Now let  $b \rightarrow \infty$ , so  $I^2 = \lim \int_{D_b} e^{-(x^2+y^2)} dx dy = \pi$ . Therefore,  $I = \sqrt{\pi}$ . Summary:  $\int_{-\infty}^{\infty} e^{-x^2} dx = \sqrt{\pi}$ .

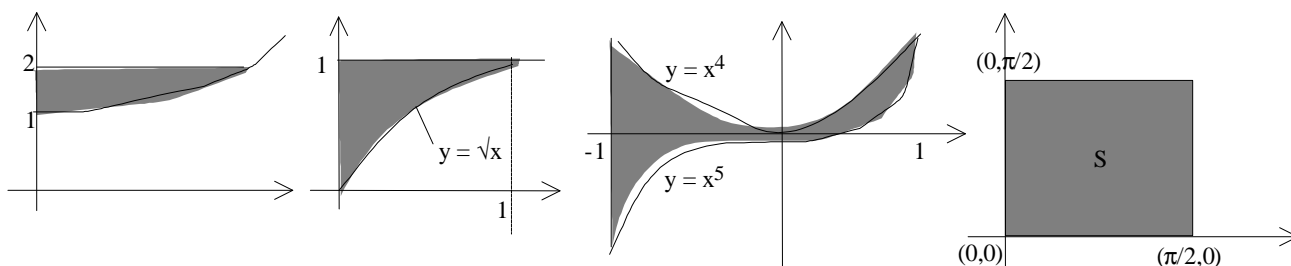
30th November 1999

### Section 3: Constrained Optimisation

This is where we *maximise* or *minimise* a function  $f$  subject to some constraints. If we have a general function  $f$  with no constraints, we could use the **Hessian** or the **steepest ascent** methods. If we have a linear function  $f$  with linear constraints, we typically use linear programming (G2M85). If we want to optimise  $(x, y, z)$  subject to a constraint of the form  $g(x, y, z) = 0$ , we handle by the method of *Lagrange multipliers*. Example: find the **optima** of  $x^3 + 15xy^{24} - 16zxy$  on the surface of  $x^2 + y^2 + z^2 - 1 = 0$ . Method: Find *values* of  $x, y, z$  and  $\lambda$  that **simultaneously** satisfy the equations  $\nabla f + \lambda \nabla g = 0$  and  $g(x, y, z) = 0$ . (Variants with  $n$ -variables and  $m$  constraints exist). The **first** condition is interpreted as  $\frac{\partial f}{\partial x} = -\lambda \frac{\partial g}{\partial x}$ ;  $\frac{\partial f}{\partial y} = -\lambda \frac{\partial g}{\partial y}$ ;  $\frac{\partial f}{\partial z} = -\lambda \frac{\partial g}{\partial z}$ . Now we solve, remembering that  $g(x, y, z) = 0$ .

### Assignment 3

**Q:** For the following integrals, sketch the region and *evaluate* the integral. (If necessary, change the **order** of integration). (i)  $\int_1^2 \int_0^{\ln y} e^{-x} dx dy$ ; (ii)  $\int_0^1 \int_{\sqrt{x}}^1 \sin(y^{3+1/2}) dy dx$ ; (iii)  $\int_S (x+y) dx dy$ , where  $S = \{(x, y) : -1 \leq x \leq 1, y \text{ lies between } x^4 \text{ and } x^5\}$ ; (iv)  $\int_0^{\pi/2} \int_0^{\pi/2} \cos(x+y) dx dy$ .

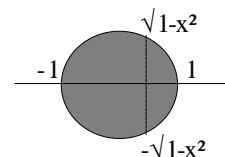


A: (i) The **lower**  $x$ -limit is  $x = \ln y$ , so  $y = e^x$ . There is no need to change the order of integration. So  $\int_1^2 \int_{\ln y}^0 e^{-x} dx dy = \int_1^2 [-e^{-x}]_{\ln y}^0 dy = \int_1^2 \{1 - 1/y\} dy = [y - \ln y]_1^2 = 2 - \ln 2 - 1 + \ln 1 = 1 - \ln 2$ . (ii) We need to *change* the order. Note that if  $y = \sqrt{x}$ , then  $x = y^2$ . So  $\int_0^1 \int_{\sqrt{x}}^1 \sin(y^{3+1/2}) dy dx = \int_0^1 \int_0^{y^2} \sin(y^{3+1/2}) dx dy = \int_0^1 [x \sin(y^{3+1/2})]_0^{y^2} dy = \int_0^1 y^2 \sin(y^{3+1/2}) dy = [-2/3 \cos(y^{3+1/2})]_0^1 = 2/3(\cos^{1/2} - \cos 1)$ .

(iii) Within the  $x$ -range  $(-1 \leq x \leq 1)$ , we have  $x^4 \geq x^5$ , so no problems.  $\int_0^1 \int_{x^4}^{x^5} (x+y) dx dy = \int_{-1}^1 \int_{x^4}^{x^5} (x+y) dy dx = \int_{-1}^1 [xy + y^2/2]_{x^4}^{x^5} dx = \int_{-1}^1 (x^5 + (x^8/2) - x^6 - (x^{10}/2)) dx = [(x^6/6) + (x^9/18) - (x^7/7) - (x^{11}/22)]_{-1}^1 = 1/9 - 2/7 - 1/11 = -184/693$ . (iv)  $\int_0^{\pi/2} \int_0^{\pi/2} \cos(x+y) dx dy = \int_0^{\pi/2} [\sin(x+y)]_{\pi/2}^0 dy = \int_0^{\pi/2} (\sin(y + \pi/2) - \sin(y)) dy = [-\cos(y + \pi/2) + \cos(y)]_{\pi/2}^0 = -\cos(\pi/2 + \pi/2) + \cos(\pi/2) + \cos(\pi/2) - \cos(0) = 0$ .

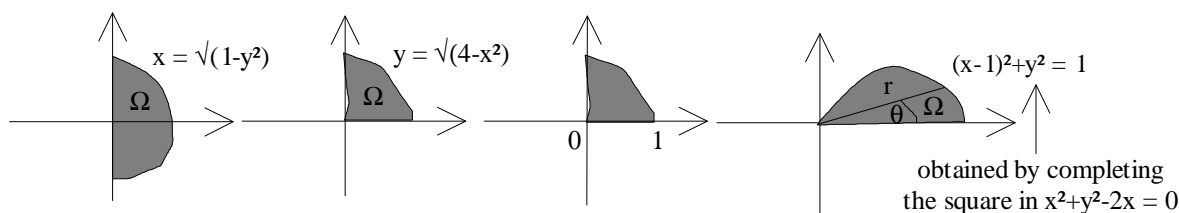
Q: Find the **volume** of the paraboloid  $z = x^2 + y^2$  under the cylinder  $x^2 + y^2 \leq 1$ .

A: The required integral is  $\int_{-1}^1 \int_{-\sqrt{1-x^2}}^{\sqrt{1-x^2}} (x^2 + y^2) dy dx = \int_{-1}^1 (x^2 y + y^3/3)_{-\sqrt{1-x^2}}^{\sqrt{1-x^2}} dx = 2 \int_{-1}^1 (x^2(1-x^2)^{1/2} + 1/3(1-x^2)^{3/2}) dx = \int_{-1}^1 (1-x^2)^{1/2} (2x^2 + 2/3(1-x^2)) dx = \int_{-1}^1 2/3(1-x^2)^{1/2} dx + 4/3 \int_{-1}^1 x^2(1-x^2) dx$ . To evaluate this either plough on, or use *substitutions*, polars, *volumes* of revolution, etc.



## Tutorial

For each of the following, change the "*Cartesian* integral" to an equivalent "*polar* integral" and then evaluate. (i)  $\int_{-1}^1 \int_0^{\sqrt{1-y^2}} \sqrt{x^2 + y^2} dx dy$ ; (ii)  $\int_0^2 \int_0^{\sqrt{4-x^2}} \sqrt{x^2 + y^2} dy dx$ ; (iii)  $\int_0^1 \int_0^{\sqrt{1-x^2}} \sin(x^2 + y^2) dy dx$ ; (iv)  $\int_0^2 \int_0^{\sqrt{2x-x^2}} x dy dx$ .



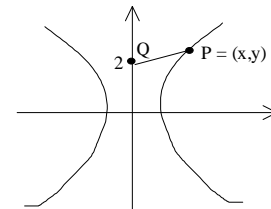
A: (i) Let  $I = \int_{-1}^1 \int_0^{\sqrt{1-y^2}} \sqrt{x^2 + y^2} dx dy$ . If  $x = r \cos \theta$  and  $y = r \sin \theta$  (*polars*), then  $I = \int_0^{\pi/2} \int_0^{\pi/2} \sqrt{(r^2 \cos^2 \theta + r^2 \sin^2 \theta)} \cdot r dr d\theta = \int_0^{\pi/2} \int_0^{\pi/2} r \cdot r dr d\theta$  (with  $\{(r, \theta) \mid 0 \leq r \leq 1, -\pi/2 \leq \theta \leq \pi/2\}$ )  $= \int_0^{\pi/2} (\int_0^{\pi/2} r^2 dr) d\theta = \pi/3$ . (ii)  $\int_0^2 \int_0^{\sqrt{4-x^2}} \sqrt{x^2 + y^2} dy dx = \int_0^{\pi/2} \int_0^2 r^2 dr d\theta = \dots = 4/3 \pi^3$ . (iii)  $\int_0^1 \int_0^{\sqrt{1-x^2}} \sin(x^2 + y^2) dy dx = \int_0^{\pi/2} \int_0^1 \sin(r^2) \cdot r dr d\theta = \pi/4 (1 - \cos 1)$ . (iv) Let  $I = \int_0^2 \int_0^{\sqrt{2x-x^2}} x dy dx$ . There are several ways to answer this question; one is to let  $x = 1 + r \cos \theta$  and  $y = r \sin \theta$ . So  $I = \int_0^{\pi} \int_0^1 (1 + r \cos \theta) r dr d\theta$  (Do check the *Jacobian*)  $= \dots = \int_0^{\pi} [1/2 - 1/3 \cos \theta] d\theta = \dots$

▶ 6th December 1999

Example: Maximise the volume of a **box** with fixed surface area. So we maximise  $f(x, y, z) = xyz$  subject to  $2(xy + yz + xz) = A$ . Set  $g(x, y, z) = xy + yz + xz - A/2$ . (So  $g = 0$  is the *constraint*). There are **four** equations:  $yz = -\lambda(y+z)$ ;  $xz = -\lambda(x+z)$ ;  $xy = -\lambda(x+y)$ ; and  $xy + yz + xz = A/2$ . So we have  $-\lambda = yz/(y+z) = xz/(x+z) = xy/(x+y)$ .

Therefore,  $yz(x+z) = xz(y+z)$ , i.e.  $yz^2 = xz^2$ , or  $z^2(x-y) = 0$ . The *stationary points* will imply that  $z = 0$  (which will not give us a **maximum**) or  $x = y$ . Symmetrically, we also have  $x = z$  and  $y = z$ . The constraint then gives  $3x^2 = A/2$ ;  $x = \sqrt{A/6}$ . (The -ve square root has no *relevance* here). So  $V_{\max} = (A/6)^{3/2}$ .

**Example:**  $x^2 - y^2 = 1$ , a *hyperbola*. What is the shortest distance to the hyperbola from the point  $(0,2)$ ? A: Here,  $f(x,y) = x^2 + (y-2)^2$ , and  $g(x,y) = x^2 - y^2 - 1$ . Note that  $x \neq 0$  because  $g(x,y) \neq 0$ . The **Lagrange** conditions give  $2x = -2\lambda x$  and  $2(y-2) = 2\lambda y$ . From  $2x = -2\lambda x$ ,  $\lambda = -1$ . So  $2(y-2) = -2y$ ;  $4y = 4$ , implying that  $y = 1$ . Substituting into  $g$  gives  $x^2 - 1 - 1 = 0$ ;  $x = \pm\sqrt{2}$ , and  $f(\pm\sqrt{2}, 1) = 3$ . Exercise: Find the **critical** points of  $f(x,y,z) = xyz$  subject to  $x^2 + y^2 + z^2$ , and *classify* them (there are lots of them).



7th December 1999

## Hessian Methods for Constrained Optimisation

Let  $l = f(x,y) + \lambda g(x,y)$ . Find the *partials* of  $l$  w.r.t.  $x$ ,  $y$  and  $\lambda$ ; and equate to **zero**. So we have  $\frac{\partial f}{\partial x} + \lambda \frac{\partial g}{\partial x} = 0$  and  $\frac{\partial f}{\partial y} + \lambda \frac{\partial g}{\partial y} = 0$ , with  $g(x,y) = 0$ . Solve for  $x$  and  $y$ . The **maximisation/minimisation** of  $f$ , subject to  $g(x,y) = 0$ , is the same as *unconstrained optimisation* of  $l(x,y)$ . The second order conditions come from the **Bordered Hessian** of  $f$  and  $g$ , which is shown on the right. Evaluate this at the critical points.

$$\begin{pmatrix} \frac{\partial^2 f}{\partial x^2} & \frac{\partial^2 f}{\partial x \partial y} & \frac{\partial g}{\partial x} \\ \frac{\partial^2 f}{\partial x \partial y} & \frac{\partial^2 f}{\partial y^2} & \frac{\partial g}{\partial y} \\ \frac{\partial g}{\partial x} & \frac{\partial g}{\partial y} & 0 \end{pmatrix}$$

If  $\det(H) < 0$ , then  $f$  has a constrained **minimum** at the critical point. If  $\det(H) > 0$ ,  $f$  has a constrained **maximum**. Note that there are “*side conditions*” on  $f$  and  $g$  to get them to work. So we will only use it on simple examples. Example: optimise  $f(x,y,z) = 4x^2 + 3xy + 6y^2$  such that  $x + y - 56 = 0$ .

**Here**,  $l = 4x^2 + 3xy + 6y^2 + \lambda(x + y - 56)$ , etc., giving  $8x + 3y + \lambda = 0$ ,  $3x + 12y + \lambda = 0$ , and  $x + y = 56$  — which imply that  $x^* = 36$  and  $y^* = 20$ . (The Solutions). The Bordered Hessian is  $H = \begin{pmatrix} 8 & 3 & 0 \\ 3 & 12 & 0 \\ 0 & 0 & 1 \end{pmatrix}$ . The determinant is  $\det(H) = -14 < 0$ , so we have a *minimum*. Another example: optimise  $f(x,y) = 4x^2 - 2xy + 6y^2$  subject to  $x + y = 72$ . Here,  $l = f + \lambda g$ , and  $x^* = 42$ ;  $y^* = 30$ .

In the above, **note** that  $\lambda = -276$  and the value of  $l_{opt}$  is 9936. If we increase the *constant term* of  $g$  by  $\epsilon$ , (e.g.  $x + y = 72 + \epsilon$ ,  $\epsilon$  small), then  $l$  increases by  $273\epsilon$ . We have the **General Property of Lagrange Multipliers**:  $\lambda$  is the change in the *value* of the maximum when a small unit change in the **constraint** is made.

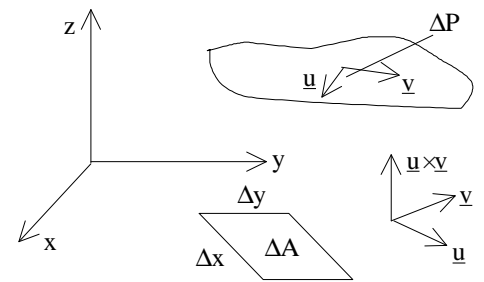
Note: this is *interpreted* as saying the following: “ $\lambda^*$  measures a property related to the **sensitivity** of the optimum value of  $f$  to *small* changes in  $g$ ”. More investigation of this will be required to get any more detail.

## Section 4: Surfaces & Surface Integrals

Let  $S$  be a surface in space, *perhaps* represented in the form  $z = f(x,y)$ , (---(1))  $[(x,y) \in \mathbf{R} \subseteq \mathbf{R}^2]$ , or by an *implicit* function  $F(x,y,z) = 0$  (---(2)), e.g.  $x^2/a^2 + y^2/b^2 + z^2/c^2 - 1 = 0$  (an *ellipsoid*), or in parametric form:  $x = f(u,v)$ ,  $y = g(u,v)$ ,  $z = h(u,v)$ . (---(3))  $[(u,v) \in \mathbf{R} \subseteq \mathbf{R}^2]$ . The area in form (1) is  $A = \iint_{R_{x,y}} \sqrt{1 + \left(\frac{\partial z}{\partial x}\right)^2 + \left(\frac{\partial z}{\partial y}\right)^2} dydx$ , and we have similar expressions in the other cases.

Note: the lecture material from now on is not examinable.

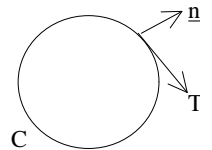
Let  $F(x,y,z) = 0$ . The area of a **surface** in a small plane is  $\approx |\underline{u} \times \underline{v}|$ , where  $\underline{u}$  and  $\underline{v}$  are tangent vectors:  $\underline{u} \approx \frac{\partial F}{\partial x} \Delta x \mathbf{e}_1$ ,  $\underline{v} \approx \frac{\partial F}{\partial y} \Delta y \mathbf{e}_2$ . The area  $\Delta P$  above  $\Delta A$  is  $\frac{\Delta A}{|\cos \gamma|}$ , where  $\gamma$  is the angle between  $\nabla F$  and the  $z$ -axis. **Therefore**,  $|\cos \gamma| = \frac{|\nabla F \cdot \underline{p}|}{\|\nabla F\|}$  where  $\underline{p} = \mathbf{e}_3$ , the unit vector in the  $z$  direction. So the *surface* area is  $\iint_{R_{xy}} \frac{\|\nabla F\|}{|\nabla F \cdot \underline{p}|} dx dy$ , provided we have  $\nabla F \cdot \underline{p} \neq 0$ .



Example: Let  $F(x,y,z) = x^2+y^2-z$ . (A paraboloid). Then  $\nabla F = (2x, 2y, -1)$ ;  $|\nabla F| = \sqrt{4x^2+4y^2+1}$ ;  $|\nabla F \cdot \underline{p}| = |-1| = 1$ ; and  $S = \iint_R \sqrt{4x^2+4y^2+1} dx dy$ . (Compare with the *other* formula). **Surface Integrals.** Let  $g$  be defined by  $g: S \rightarrow \mathbf{R}$ , where  $S$  is given by  $F(x,y,z) = 0$ . The Surface Integral is  $\iint_S g dA = \iint_{R_{xy}} g(x,y,z) \frac{\|\nabla F\|}{|\nabla F \cdot \underline{p}|} dx dy$ . **Vector Fields.** Let  $F$  be defined by  $F: S \rightarrow \mathbf{R}^3$ , where  $S \subset \mathbf{R}^3$ . So  $F = M\underline{i} + N\underline{j} + P\underline{k}$ . Example: If  $F: S \rightarrow \mathbf{R}^3$  is differentiable, then  $\nabla F = (\frac{\partial f}{\partial x}, \frac{\partial f}{\partial y}, \frac{\partial f}{\partial z})$  is the *gradient vector field* of  $F$  on  $S$ .

## Flux

Consider Flux across a *plane curve*. Let  $C$  be a smooth closed curve, defined by  $x = g(t)$  and  $y = h(t)$ , with  $t \in [a,b]$ . The tangent vector  $T$  is defined by  $(\frac{dx}{ds}\underline{i} + \frac{dy}{ds}\underline{j})$ . Take  $\underline{k}$ , a *unit vector* in the  $z$  direction. The outward unit normal  $\underline{n}$ , of  $C$ , is defined by  $\underline{n} = \underline{T} \times \underline{k} = \frac{dy}{ds}\underline{i} - \frac{dx}{ds}\underline{j}$ . If  $f: S \rightarrow \mathbf{R}^3$  is a vector field with  $C \subset S$ , the *flux of f across C* is given by the formula  $\int_C \underline{F} \cdot \underline{n} = \int_C (M \frac{dy}{ds} - N \frac{dx}{ds}) ds$ .



Example: Let  $\underline{F} = (x-y)\underline{i} + x\underline{j}$ , and let  $C = \{x^2+y^2=1\}$ . **Parametrise:**  $r = \cos t \underline{i} + \sin t \underline{j}$ , for  $0 \leq t \leq 2\pi$ ;  $M = \cos t - \sin t$  ( $= x-y$ );  $N = \cos t$ . So  $\frac{dy}{dt} = \cos t$  and  $\frac{dx}{dt} = -\sin t$ . **Flux:**  $\int_C (M \frac{dy}{ds} - N \frac{dx}{ds}) ds = \int_0^{2\pi} (\cos^2 t - \sin t \cos t) dt = \dots = \pi$ . Note that  $\pi > 0$ , so the net flux is *outward*.

## Green's Theorem

Green's Theorem (in the **plane**):  $\int_C \underline{F} \cdot \underline{n} dS = \iint_R (\frac{\partial M}{\partial x} + \frac{\partial N}{\partial y}) dx dy$  (the region *enclosed* by  $C$ ). There is a 3 dimensional version: given a surface  $S$  in  $\mathbf{R}^3$ , assume that we can choose a *unit normal vector field*  $\underline{n}$  ( $S$  can be **oriented**). The *flux* across  $S$  is given by  $\int_S \underline{F} \cdot \underline{n} d\sigma$ , where  $d\sigma = \frac{\|\nabla f\|}{|\nabla f \cdot \underline{p}|} dA$  as *before*, with  $S = \{(x,y,z) \mid f(x,y,z) = 0\}$ .

Example:  $F(x,y,z) = x\underline{i} + y\underline{j}$ . Calculate the flux of  $F$  **across** the surface  $x^2+y^2+z^2 = a^2$  ( $\underline{n}$  is outward). The *outward normal vector field*  $\underline{n}$  is given by the gradient  $\nabla f$ , where  $f(x,y,z) = x^2+y^2+z^2-a^2$  after *normalisation*. So  $\nabla f = (2x, 2y, 2z)$ , and points outwards (good!). It follows that  $\underline{n} = \frac{\nabla f}{\|\nabla f\|} = \frac{(2x, 2y, 2z)}{\sqrt{4x^2+4y^2+4z^2}} = \frac{(2x, 2y, 2z)}{\sqrt{4a^2}} = (\frac{x}{a}, \frac{y}{a}, \frac{z}{a})$ .

**Now**  $\underline{F} \cdot \underline{n} = \frac{1}{a}(x,y,0) \cdot (x,y,z) = \frac{1}{a}(x^2+y^2)$ . The flux *out* of  $S$  is  $\frac{1}{a} \int_S (x^2+y^2) d\sigma$ . The usual *parametrisation* of  $S$  is  $r(u,v) = (a \cos u \cos v, a \sin u \cos v, a \sin v)$ . ( $0 \leq u \leq 2\pi$ ), ( $-\frac{\pi}{2} \leq v \leq \frac{\pi}{2}$ ), etc.

# Stoke's Theorem (in R<sup>3</sup>)

Given C, a *boundary curve* of an orientable surface S, then  $\int_C \underline{F} \cdot d\underline{r} = \int_S (\underline{\nabla} \times \underline{F}) \cdot \underline{n} \, d\sigma$ .

## Gauss' Divergence Theorem

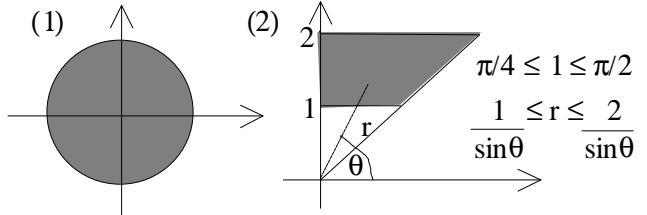
$\int_S \underline{F} \cdot \underline{n} \, d\sigma = \int_D \underline{\nabla} \cdot \underline{F} \, dV$  (*The Boundary* of D is S). Here,  $\underline{F}$  is a vector field across the *boundary* of the region D. Notes:  $\underline{\nabla} \cdot \underline{F} = (\frac{\partial}{\partial x}, \frac{\partial}{\partial y}, \frac{\partial}{\partial z}) \cdot (M, N, P) = \frac{\partial}{\partial x}M + \frac{\partial}{\partial y}N + \frac{\partial}{\partial z}P$ . And so

$$\underline{\nabla} \times \underline{F} = \begin{vmatrix} \mathbf{i} & \mathbf{j} & \mathbf{k} \\ \frac{\partial}{\partial x} & \frac{\partial}{\partial y} & \frac{\partial}{\partial z} \\ M & N & P \end{vmatrix}.$$

A matrix  $(A \ B \ C)$  is positive semi-definite if  $A \geq 0$ ,  $C \geq 0$ , and  $AC - B^2 \geq 0$ . **Similar** conditions apply for negative definite, etc.).

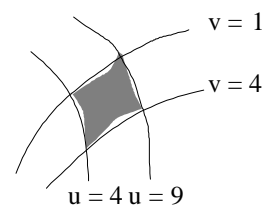
## Assignment 4

Q: By changing to a *polar integral*, evaluate  $I = \int_{-1}^1 \int_{-\sqrt{1-x^2}}^{\sqrt{1-x^2}} e^{-(x^2+y^2)} dy dx$ . A:  $I = \int_{\theta=0}^{2\pi} \int_{r=0}^1 e^{-r^2} r dr d\theta = 2\pi [e^{-r^2}/2]_0^1 = \pi(e^{-1}-1)$ . Q: Evaluate  $I = \int_{\theta=\pi/4}^{\pi/2} \int_{r=1/\sin\theta}^{2/\sin\theta} r \cos\theta / \sin\theta \, r dr d\theta$ . A:  $I = \int_{\pi/4}^{\pi/2} \frac{\cos\theta}{\sin\theta} [\frac{r^3}{3}]_{1/\sin\theta}^{2/\sin\theta} d\theta = \frac{1}{3} \int_{\pi/4}^{\pi/2} \frac{\cos\theta}{\sin^3\theta} (8 - 1) d\theta = \frac{7}{9} \int_{\pi/4}^{\pi/2} \frac{\cos\theta}{\sin^3\theta} d\theta = \frac{7}{9} [-\frac{1}{2\sin^2\theta}]_{\pi/4}^{\pi/2} = \frac{7}{9}(1 - 2\sqrt{2}) = \frac{7}{9}(2\sqrt{2}-1)$ .



Q: Evaluate the following using *cylindrical co-ordinates*:  $I = \int_0^1 \int_0^{\sqrt{1-x^2}} \int_0^{2\sqrt{1-x^2-y^2}} \cos(x^2+y^2) dz dy dx$ . A:  $I = 2 \int_0^1 \int_0^{\sqrt{1-x^2}} \cos(x^2+y^2) dy dx = 2 \int_0^{\pi/2} \int_0^1 r \cos(r^2) dr d\theta = \pi [\sin(r^2)]_0^1 = \pi \sin(1)$ .

Q: Evaluate  $\int_{\Omega} xy dx dy$ , where  $\Omega$  is the region in the *1st quadrant* bounded by the curves  $x^2+y^2 = 4$ ,  $x^2+y^2 = 9$ ,  $x^2-y^2 = 1$  and  $x^2-y^2 = 4$ . (Use the **transition**  $u = x^2+y^2$  and  $v = x^2-y^2$ ). A: We know that  $u+v = 2x^2$  and  $u-v = 2y^2$  so  $x = \sqrt{(u+v)/2}$  and  $y = \sqrt{(u-v)/2}$ . So  $J(u,v) = |\frac{\partial x}{\partial u} \frac{\partial y}{\partial v} - \frac{\partial x}{\partial v} \frac{\partial y}{\partial u}|$ . Calculate this *directly* ( $\frac{\partial x}{\partial u} = \frac{1}{2\sqrt{2(u+v)^{3/2}}$ ,  $\frac{\partial x}{\partial v} = \frac{1}{2\sqrt{2(u-v)^{3/2}}$ ) to get  $J(u,v) = \frac{1}{4\sqrt{(u^2-v^2)}}$ . Therefore, the *integral* is  $\int_4^9 \int_{u-v}^{u+v} \frac{1}{4\sqrt{(u^2-v^2)}} \sqrt{(u^2-v^2)/2} \times \frac{1}{4\sqrt{(u^2-v^2)}} du dv = \frac{1}{8} \int_4^9 \int_{u-v}^{u+v} du dv = \frac{15}{8}$ .



# Exam Paper: January 2000

## SECTION 1 (Compulsory)

- (1) (a) Calculate the Hessian of the function:  
 $f(x, y) = -xye^{-(x^2+y^2)/2}$ .  
Find and classify all the critical points of  $f$ . **[12 marks]**
- (b) The volume  $V$  of a sphere of radius  $R$  is given in cartesian coordinates by  
 $V = 2 \iint_{\Omega} \sqrt{R^2 - (x^2 + y^2)} \, dx dy$ ,  
where  $\Omega$  is the disc of radius  $R$  centred at the origin. Use polar coordinates to show  
that  $V = \frac{4}{3}\pi R^3$ . **[8 marks]**

## SECTION 2 (Answer 2 out of 4 questions)

- (2) By using the method of Lagrange multipliers, minimise  $xyz$  subject to the constraint  
 $8x^3 - 8y^3 + 27z^3 = 1$ , with  $x \geq 0$ , and  $z \geq 0$ , whilst  $y \leq 0$ . **[15 marks]**
- (3) (a) Evaluate the triple integral  $\int_0^2 \int_0^x \int_0^{4-x^2} xyz \, dz \, dy \, dx$ . **[8 marks]**  
(b) Use the transformation  $u = x+y$ ,  $v = 2x-y$  to simplify the integral  $\iint_{\Omega} (x+y)^2 dx dy$   
where  $\Omega$  is the parallelogram bounded by the lines  $x+y=0$ ,  $x+y=1$ ,  $2x-y=0$ ,  $2x-y=3$ .  
**[7 marks]**
- (4) (a) Determine whether or not the following function is continuous at the origin. If the  
function is continuous then prove it, otherwise provide reasons for the lack of  
continuity:  $f(x, y) = \frac{2xy}{x^2+y^2}$ , if  $(x, y) \neq (0, 0)$ ;  $f(0, 0) = 0$ . **[6 marks]**
- (b) A function  $f: S \rightarrow \mathbb{R}$  is defined on a convex subset  $S$  of  $\mathbb{R}^n$ . You are told that at some  
point  $\mathbf{x}_0 \in S$ ,  $f$  has a local minimum (so there is some  $\epsilon > 0$  such that if  $\|\mathbf{x} - \mathbf{x}_0\| < \epsilon$   
then  $f(\mathbf{x}) \geq f(\mathbf{x}_0)$ .) Supposing that  $f$  is a convex function on  $S$ , prove that  $f$  has a  
global minimum at  $\mathbf{x}_0$ . **[9 marks]**
- (5) Consider the mapping given by  $u = e^x \cosh(y)$ ,  $v = e^x \sinh(y)$ .
- (i) Find the Jacobian  $\frac{\partial(u, v)}{\partial(x, y)}$  of the transformation.  
Determine where local inverses exist. **[6 marks]**
- (ii) Solve for  $x, y$  in terms of  $u, v$  to find the inverse mapping and calculate its Jacobian,  
giving your answer in terms of  $u, v$  and also in terms of  $x, y$ :  $\frac{\partial(x, y)}{\partial(u, v)}$ .  
**[6 marks]**

Finally verify  $\frac{\partial(u, v)}{\partial(x, y)} \frac{\partial(x, y)}{\partial(u, v)} = \mathbf{I}$ , the two-by-two identity matrix.

**[3 marks]**

(Questions done: 1, 3, 5)