EE2 Maths Summary Sheet: Functions of Multiple Variables

1) The change in f following a small step $\Delta x, \Delta y$ is:

$$\Delta f \approx \frac{\partial f(x,y)}{\partial x} \Delta x + \frac{\partial f(x,y)}{\partial y} \Delta y.$$
(1)

The error in this approximation goes to zero as $\Delta x, \Delta y \rightarrow 0$.

2) In the limit $\Delta x, \Delta y \to 0$ Eq. (1) becomes the **Total Differential**:

$$df = \frac{\partial f(x,y)}{\partial x} dx + \frac{\partial f(x,y)}{\partial y} dy$$
(2)

This generalises to $df = \frac{\partial f}{\partial x_1} dx_1 + \frac{\partial f}{\partial x_2} dx_2 + \dots \frac{\partial f}{\partial x_n} dx_n$ for functions of n variables.

3) Sometimes we have f(x, y) but x and y are both themselves functions of a variable u such that x(u), y(u). In this case the **Chain Rule** applies:

$$\frac{df}{du} = \frac{\partial f(x,y)}{\partial x} \frac{dx}{du} + \frac{\partial f(x,y)}{\partial y} \frac{dy}{du}.$$
(3)

Note that this only applies when f is a univariate function of u.

4) Sometimes we can express f(x, y) in another co-ordinate system as f(u, v). In this case we can relate the co-ordinate systems by x = x(u, v) and y = y(u, v). We can then relate the partial differentials of f as follows:

$$\frac{\partial f}{\partial u}\Big|_{v} = \frac{\partial f}{\partial x}\Big|_{y} \frac{\partial x}{\partial u}\Big|_{v} + \frac{\partial f}{\partial y}\Big|_{x} \frac{\partial y}{\partial u}\Big|_{v}$$
(4)

$$\frac{\partial f}{\partial v}\Big|_{u} = \frac{\partial f}{\partial x}\Big|_{y}\frac{\partial x}{\partial v}\Big|_{u} + \frac{\partial f}{\partial y}\Big|_{x}\frac{\partial y}{\partial v}\Big|_{u}$$
(5)

A rule of thumb to obtain e.g. Eq. (5) is to take the total differential Eq. (2) and "multiply through by $\frac{\partial v}{\partial v}\Big|_{u}$ ". Equations like Eq. (4,5) are sometimes referred to as "expressing a change of variables in f".

5) A Taylor's series expansion about a point can be generalized to the multivariate case. Below is the expression for an expansion of f(x, y) about (x_0, y_0) where we define $\Delta x = x - x_0$ and $\Delta y = y - y_0$ and evaluate the derivatives at x_0 , y_0 :

$$f(x,y) = f(x_0,y_0) + \Delta x \frac{\partial f}{\partial x} + \Delta y \frac{\partial f}{\partial y} + \frac{1}{2!} \left[\frac{\partial^2 f}{\partial x^2} (\Delta x)^2 + 2 \frac{\partial^2 f}{\partial x \partial y} \Delta x \Delta y + \frac{\partial^2 f}{\partial y^2} (\Delta y)^2 \right] + \dots$$
(6)

6) The total differential Eq. (2) can be written as $df = d\underline{s} \cdot \nabla f$ where $d\underline{s} = (dx, dy)$ and $\nabla f = (\frac{\partial f}{\partial x}, \frac{\partial f}{\partial y})$ is called the *Gradient Vector*. ∇f : points, in the x-y-plane, in the direction of greatest change of f. It's magnitude $|\nabla f(x, y)|$ tells us the gradient in the direction $\nabla f(x, y)$.

7) There are three types of stationary points of f(x, y): Maxima, Minima and Saddle Points. The sufficient conditions for a stationary point to be a Max, Min, Saddle are:

 $Max: \quad f_{xx} < 0 \quad \text{and} \quad f_{xy}^2 - f_{xx}f_{yy} < 0 \tag{7}$

:
$$f_{xx} > 0$$
 and $f_{xy}^2 - f_{xx}f_{yy} < 0$ (8)

$$Saddle: \quad f_{xy}^2 - f_{xx}f_{yy} > 0. \tag{9}$$

8) If $f(x_1...x_n)$ we can define $H_{ij} = \frac{\partial}{\partial x_i} \frac{\partial}{\partial x_j} f(x_1, ..., x_n)$. The matrix H is called the *Hessian*. 9) Given that

$$F(x) = \int_{t=u(x)}^{t=v(x)} f(x,t)dt$$
(10)

the Leibnitz' Integral Rule is then:

Min

$$\frac{dF(x)}{dx} = f(x, v(x))\frac{dv}{dx} - f(x, u(x))\frac{du}{dx} + \int_{t=u(x)}^{t=v(x)} \frac{\partial f}{\partial x}dt.$$
(11)