

Math 237 Reference

Definition 1 *Scalar Function* A scalar function $f(x_1, \dots, x_n)$ of n variables is a function whose domain is a subset of \mathbb{R}^n and whose range is a subset of \mathbb{R} .

Definition 2 *Level Curves* The level curves of a function $f(x, y)$ are the curves

$$f(x, y) = k$$

where $k \in R(f)$.

Definition 3 *Level Surfaces* A level surface of a scalar function $f(x, y, z)$ is defined by

$$f(x, y, z) = k$$

where $k \in R(f)$.

Definition 4 *Level Sets* A level set of a scalar function $f(\vec{x})$, $\vec{x} \in \mathbb{R}^n$ is defined by

$$\{\vec{x} \in \mathbb{R}^n : f(\vec{x}) = k, k \in R(f)\}$$

Definition 5 *Neighbourhood* An r -neighbourhood of a point $(a, b) \in \mathbb{R}^2$ is a set

$$N_r(a, b) = \{(x, y) \in \mathbb{R}^2 : \|(x, y) - (a, b)\| < r, r \in \mathbb{R}\}$$

Definition 6 *Limit* Assume $f(x, y)$ is defined in a neighbourhood of (a, b) , except possibly at (a, b) . If, for every $\epsilon > 0$, there exists a $\delta > 0$ such that

$$0 < \|(x, y) - (a, b)\| < \delta \Rightarrow |f(x, y) - L| < \epsilon$$

then

$$\lim_{(x, y) \rightarrow (a, b)} f(x, y) = L$$

Theorem 1 *Limits 1* If $\lim_{(x, y) \rightarrow (a, b)} f(x, y)$ and $\lim_{(x, y) \rightarrow (a, b)} g(x, y)$ both exist,

1. $\lim_{(x, y) \rightarrow (a, b)} [f(x, y) + g(x, y)] = \lim_{(x, y) \rightarrow (a, b)} f(x, y) + \lim_{(x, y) \rightarrow (a, b)} g(x, y)$
2. $\lim_{(x, y) \rightarrow (a, b)} [f(x, y)g(x, y)] = \left[\lim_{(x, y) \rightarrow (a, b)} f(x, y) \right] \left[\lim_{(x, y) \rightarrow (a, b)} g(x, y) \right]$
3. $\lim_{(x, y) \rightarrow (a, b)} \frac{f(x, y)}{g(x, y)} = \frac{\lim_{(x, y) \rightarrow (a, b)} f(x, y)}{\lim_{(x, y) \rightarrow (a, b)} g(x, y)}$ provided $\lim_{(x, y) \rightarrow (a, b)} g(x, y) \neq 0$

Theorem 2 *Limits 2* If $\lim_{(x, y) \rightarrow (a, b)} f(x, y)$ exists, then the limit is unique.

Theorem 3 Squeeze Theorem If there exists a function $B(x, y)$ such that

$$|f(x, y) - L| \leq B(x, y) \text{ for all } (x, y) \neq (a, b)$$

in some neighbourhood of (a, b) and $\lim_{(x, y) \rightarrow (a, b)} B(x, y) = 0$, then

$$\lim_{(x, y) \rightarrow (a, b)} f(x, y) = L$$

Theorem 4 Continuity 1 If f and g are both continuous at (a, b) , then $f + g$ and fg are continuous at (a, b) .

Theorem 5 Continuity 2 If f and g are both continuous at (a, b) and $g(a, b) \neq 0$, then the quotient $\frac{f}{g}$ is continuous at (a, b) .

Theorem 6 Continuity 3 If $f(x, y)$ is continuous at (a, b) and $g(t)$ is continuous at $f(a, b)$, then the composition $g \circ f$ is continuous at (a, b) .

Definition 7 Partial Derivatives The partial derivatives of $f(x, y)$ are defined by

$$\frac{\partial f}{\partial x}(x, y) = f_x(x, y) = \lim_{h \rightarrow 0} \frac{f(x + h, y) - f(x, y)}{h}$$

$$\frac{\partial f}{\partial y}(x, y) = f_y(x, y) = \lim_{h \rightarrow 0} \frac{f(x, y + h) - f(x, y)}{h}$$

Theorem 7 Clairaut's Theorem If f_{xy} and f_{yx} are defined in some neighbourhood of (a, b) and are both continuous at (a, b) , then

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

Definition 8 Tangent Plane The tangent plane to $z = f(x, y)$ at the point $(a, b, f(a, b))$ is

$$z = f(a, b) + \frac{\partial f}{\partial x}(a, b)(x - a) + \frac{\partial f}{\partial y}(a, b)(y - b)$$

Definition 9 Linearization and Linear Approximation For a function $f(x, y)$ we define the linearization $L_{(a, b)}(x, y)$ of f at (a, b) by

$$L_{(a, b)}(x, y) = f(a, b) + \frac{\partial f}{\partial x}(a, b)(x - a) + \frac{\partial f}{\partial y}(a, b)(y - b)$$

We call the approximation $f(x, y) \approx L_{(a, b)}(x, y)$ the linear approximation of $f(x, y)$ at (a, b) .

Definition 10 Gradient Suppose that $f(x, y, z)$ has partial derivatives at $\vec{a} \in \mathbb{R}^3$. The gradient of f at \vec{a} is defined by

$$\nabla f(\vec{a}) = (f_x(\vec{a}), f_y(\vec{a}), f_z(\vec{a}))$$

Definition 11 Linearization and Linear Approximation Suppose that $f(\vec{x})$, $\vec{x} \in \mathbb{R}^3$, has partial derivatives at $\vec{a} \in \mathbb{R}^3$. The linearization of f at \vec{a} is defined by

$$L_{\vec{a}}(\vec{x}) = f(\vec{a}) + \nabla f(\vec{a})(\vec{x} - \vec{a})$$

The linear approximation of f at \vec{a} is

$$f(\vec{x}) \approx f(\vec{a}) + \nabla f(\vec{a})(\vec{x} - \vec{a})$$

Theorem 8 Differentiability 1 If $g'(a)$ exists, then $\lim_{x \rightarrow a} \frac{|R_{1,a}(x)|}{|x - a|} = 0$ where

$$R_{1,a}(x) = g(x) - L_a(x) = g(x) - g(a) - g'(a)(x - a)$$

Definition 12 Differentiable A function $f(x, y)$ is differentiable at (a, b) if

$$\lim_{(x,y) \rightarrow (a,b)} \frac{|R_{1,(a,b)}(x, y)|}{\|(x, y) - (a, b)\|} = 0$$

where

$$R_{1,(a,b)}(x, y) = f(x, y) - L_{(a,b)}(x, y)$$

Theorem 9 Differentiability 2 If a function $f(x, y)$ satisfies

$$\lim_{(x,y) \rightarrow (a,b)} \frac{|f(x, y) - f(a, b) - c(x - a) - d(y - b)|}{\|(x, y) - (a, b)\|} = 0$$

for some constants c and d , then $c = f_x(a, b)$ and $d = f_y(a, b)$.

Theorem 10 Continuity If $f(x, y)$ is differentiable at (a, b) , then f is continuous at (a, b) .

Theorem 11 If the partial derivatives $\frac{\partial f}{\partial x}$ and $\frac{\partial f}{\partial y}$ are both continuous at (a, b) , then $f(x, y)$ is differentiable at (a, b) .

Definition 13 Differentiability for $f : \mathbb{R}^n \rightarrow \mathbb{R}$ A function $f : \mathbb{R}^n \rightarrow \mathbb{R}$ is differentiable at a point $\vec{a} = (a_1, \dots, a_n)$ if

$$\lim_{\vec{x} \rightarrow \vec{a}} \frac{|f(\vec{x}) - f(\vec{a}) - L_{\vec{a}}(\vec{x} - \vec{a})|}{\|\vec{x} - \vec{a}\|} = 0$$

where $L : \mathbb{R}^n \rightarrow \mathbb{R}$ is a linear transformation.

Theorem 12 If $f(x_1, \dots, x_n)$ is differentiable at $\vec{a} = (a_1, \dots, a_n)$, then f is continuous at \vec{a} .

Theorem 13 If $\frac{\partial f}{\partial x_1}, \dots, \frac{\partial f}{\partial x_n}$ are continuous at $\vec{a} = (a_1, \dots, a_n)$, then $f(x_1, \dots, x_n)$ is differentiable at \vec{a} .

Theorem 14 Chain Rule Let $G(t) = f(x(t), y(t))$ and let $a = x(t_0)$ and $b = y(t_0)$. If f is differentiable at (a, b) and $x'(t_0)$ and $y'(t_0)$ exist, then $G'(t_0)$ exists and is given by

$$G'(t_0) = f_x(a, b)x'(t_0) + f_y(a, b)y'(t_0)$$

Definition 14 Directional Derivative The directional derivative of $f(x, y)$ at point (a, b) in the direction of a unit vector $\vec{u} = (u_1, u_2)$ where $\|\vec{u}\| = 1$ is defined by

$$D_{\vec{u}}f(a, b) = \left. \frac{d}{ds} f(a + su_1, b + su_2) \right|_{s=0} = \lim_{t \rightarrow 0} \frac{f((a, b) + t\vec{u}) - f(a, b)}{t}$$

Theorem 15 Directional Derivative Theorem If $f(x, y)$ is differentiable at (a, b) and $\vec{u} = (u_1, u_2)$ where $\|\vec{u}\| = 1$, then

$$D_{\vec{u}}f(a, b) = \nabla f(a, b) \cdot \vec{u}$$

Theorem 16 Greatest Rate of Change If $f(x, y)$ is differentiable at (a, b) and $\nabla f(a, b) \neq (0, 0)$, then the largest value of $D_{\vec{u}}f(a, b)$ is $\|\nabla f(a, b)\|$, and occurs when \vec{u} is in the direction of $\nabla f(a, b)$.

Theorem 17 Orthogonality Theorem If $f(x, y) \in C^1$ in a neighbourhood of (a, b) and $\nabla f(a, b) \neq (0, 0)$, then $\nabla f(a, b)$ is orthogonal to the level curve $f(x, y) = k$ through (a, b) .

This theorem extends to three dimensions for $f(x, y, z)$ and (a, b, c) .

Definition 15 2nd Degree Taylor Polynomial Let f be a function of two variables. The second degree Taylor Polynomial $P_{2,(a,b)}$ of $f(x, y)$ at (a, b) is given by

$$\begin{aligned} P_{2,(a,b)}(x, y) &= f(a, b) + f_x(a, b)(x - a) + f_y(a, b)(y - b) \\ &= \frac{1}{2} [f_{xx}(a, b)(x - a)^2 + 2f_{xy}(a, b)(x - a)(y - b) + f_{yy}(a, b)(y - b)^2] \end{aligned}$$

Definition 16 Hessian Matrix The Hessian matrix of $f(x, y)$, denoted by $Hf(x, y)$, is defined as

$$\begin{bmatrix} f_{xx}(x, y) & f_{xy}(x, y) \\ f_{yx}(x, y) & f_{yy}(x, y) \end{bmatrix}$$

Theorem 18 Taylor's Theorem for Functions of 2 Variables If $f(x, y) \in C^2$ in some neighbourhood $N(a, b)$ of (a, b) , then for all $(x, y) \in N(a, b)$ there exists a point (c, d) on the line segment joining (a, b) and (x, y) such that

$$f(x, y) = f(a, b) + f_x(a, b)(x - a) + f_y(a, b)(y - b) + R_{1,(a,b)}(x, y)$$

where

$$R_{1,(a,b)}(x, y) = \frac{1}{2} [f_{xx}(c, d)(x - a)^2 + 2f_{xy}(c, d)(x - a)(y - b) + f_{yy}(c, d)(y - b)^2]$$

Corollary 18.1 If $f(x, y) \in C^2$ in some closed neighbourhood $N(a, b)$ of (a, b) , then there exists a positive constant M such that

$$|R_{1,(a,b)}(x, y)| \leq M \|(x, y) - (a, b)\|^2$$

for all $(x, y) \in N(a, b)$.

Definition 17 k -th Degree Taylor Polynomial The k -th degree Taylor polynomial of a function $f(x, y)$ is

$$P_{k,(a,b)}(x, y) = \sum_{|\alpha| \leq k} \partial^\alpha f(a, b) \frac{[(x, y) - (a, b)]^\alpha}{\alpha!}$$

Definition 18 Taylor's Theorem of Order k If $f(x, y) \in C^{k+1}$ in some neighbourhood $N(a, b)$ of (a, b) , then for all $(x, y) \in N(a, b)$ there exists a point (c, d) on the line segment between (a, b) and (x, y) such that

$$f(x, y) = P_{k,(a,b)}(x, y) + R_{k,(a,b)}(x, y)$$

where

$$R_{k,(a,b)}(x, y) = \sum_{|\alpha| \leq k+1} \partial^\alpha f(c, d) \frac{[(x, y) - (a, b)]^\alpha}{\alpha!}$$

Corollary 18.2 If $f(x, y) \in C^k$ in some neighbourhood of (a, b) , then

$$\lim_{(x,y) \rightarrow (a,b)} \frac{|f(x, y) - P_{k,(a,b)}(x, y)|}{\|(x, y) - (a, b)\|^k} = 0$$

Corollary 18.3 If $f(x, y) \in C^{k+1}$ in some closed neighbourhood $N(a, b)$ of (a, b) , then there exists a constant $M > 0$ such that

$$|f(x, y) - P_{k,(a,b)}(x, y)| \leq M \|(x, y) - (a, b)\|^{k+1}$$

for all $(x, y) \in N(a, b)$.

Definition 19 Generalized Taylor Polynomial The Taylor polynomial of degree k for functions of n variables is

$$P_{k,\vec{a}}(\vec{x}) = \sum_{|\alpha| \leq k} \partial^\alpha f(\vec{a}) \frac{(\vec{x} - \vec{a})^\alpha}{\alpha!}$$

Definition 20 Local Maximum and Minimum A point (a, b) is a local maximum point of f if $f(x, y) \leq f(a, b)$ for all (x, y) in some neighbourhood of (a, b) . A point (a, b) is a local minimum point of f if $f(x, y) \geq f(a, b)$ for all (x, y) in some neighbourhood of (a, b) .

Theorem 19 If (a, b) is a local maximum or minimum point of f , then each partial derivative is either equal to zero or does not exist.

Definition 21 Critical Point A point (a, b) in the domain of $f(x, y)$ is called a critical point of f if

$$\frac{\partial f}{\partial x}(a, b) = 0 \text{ or } \frac{\partial f}{\partial x}(a, b) \text{ does not exist}$$

and

$$\frac{\partial f}{\partial y}(a, b) = 0 \text{ or } \frac{\partial f}{\partial y}(a, b) \text{ does not exist}$$

Definition 22 Saddle Point A critical point (a, b) of $f(x, y)$ is called a saddle point of f if in every neighbourhood of (a, b) there exists points (x_1, y_1) and (x_2, y_2) such that

$$f(x_1, y_1) > f(a, b) \text{ and } f(x_2, y_2) < f(a, b)$$

Definition 23 Quadratic Form A function Q of the form

$$Q(u, v) = \begin{bmatrix} u & v \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} \\ a_{12} & a_{22} \end{bmatrix} \begin{bmatrix} u \\ v \end{bmatrix} = a_{11}u^2 + 2a_{12}uv + a_{22}v^2$$

where a_{11}, a_{12} , and a_{22} are constants, is called a quadratic form on \mathbb{R}^2 .

Definition 24 Determinant and Quadratic Forms A quadratic form $Q(u, v) = a_{11}u^2 + 2a_{12}uv + a_{22}v^2$ on \mathbb{R}^2 is

1. Positive definite if $\det(A) > 0$ and $a_{11} > 0$
2. Negative definite if $\det(A) > 0$ and $a_{11} < 0$
3. Indefinite if $\det(A) < 0$
4. Semidefinite if $\det(A) = 0$

Theorem 20 Second Partial Derivatives Test Suppose that $f(x, y) \in C^2$ in some neighbourhood of (a, b) and that

$$f_x(a, b) = 0 = f_y(a, b)$$

1. If $Hf(a, b)$ is positive definite, then (a, b) is a local minimum point of f
2. If $Hf(a, b)$ is negative definite, then (a, b) is a local maximum point of f
3. If $Hf(a, b)$ is indefinite, then (a, b) is a saddle point of f
4. If $Hf(a, b)$ is semidefinite, then the test is inconclusive

Definition 25 Convex and Strictly Convex Let $f(x, y) \in C^2$. We say that f is convex if $Hf(x, y)$ is positive semi-definite for all (x, y) and that f is strictly convex if $Hf(x, y)$ is positive definite for all (x, y) .

Theorem 21 If $f(x, y) \in C^2$ and is strictly convex, then

1. $f(x, y) > L_{(a,b)}(x, y)$ for all $(x, y) \neq (a, b)$
2. $f(a_1 + t(b_1 - a_1), a_2 + t(b_2 - a_2)) < f(a_1, a_2) + t[f(b_1, b_2) - f(a_1, a_2)]$ for $0 < t < 1$ and $(a_1, a_2) \neq (b_1, b_2)$

Theorem 22 If $f(x, y) \in C^2$ is convex, then every critical point (c, d) satisfies $f(x, y) \geq f(c, d)$ for all $(x, y) \neq (c, d)$. If $f(x, y) \in C^2$ is strictly convex and has a critical point (c, d) , then $f(x, y) > f(c, d)$ for all $(x, y) \neq (c, d)$ and f has no other critical point.

Theorem 23 Let $\begin{bmatrix} a & b \\ b & c \end{bmatrix}$ be a positive definite matrix. If $|\tilde{a} - a|$, $|\tilde{b} - b|$, and $|\tilde{c} - c|$ are sufficiently small, then $\begin{bmatrix} \tilde{a} & \tilde{b} \\ \tilde{b} & \tilde{c} \end{bmatrix}$ is positive definite.

Definition 26 Absolute Maximum and Minimum Given a function $f(x, y)$ and a set $S \subseteq \mathbb{R}^2$,

1. A point $(a, b) \in S$ is an absolute maximum point of f on S if

$$f(x, y) \leq f(a, b) \text{ for all } (x, y) \in S$$

The value $f(a, b)$ is called the absolute maximum value of f on S .

2. A point $(a, b) \in S$ is an absolute minimum point of f on S if

$$f(x, y) \geq f(a, b) \text{ for all } (x, y) \in S$$

The value $f(a, b)$ is called the absolute minimum value of f on S .

Definition 27 Bounded Set A set $S \subset \mathbb{R}^2$ is said to be bounded if it is contained in some neighbourhood of the origin.

Definition 28 Boundary Point Given a set $S \subset \mathbb{R}^2$, a point $(a, b) \in \mathbb{R}^2$ is said to be a boundary point of S if every neighbourhood of (a, b) contains at least one point in S and one point not in S .

Definition 29 Boundary of S The set $B(S)$ of all boundary points of S is called the boundary of S .

Definition 30 Closed Set A set $S \subseteq \mathbb{R}^2$ is said to be closed if S contains all of its boundary points.

Theorem 24 Extreme Value Theorem If $f(x, y)$ is continuous on a closed and bounded set $S \subseteq \mathbb{R}^2$, then there exists points $(a, b), (c, d) \in S$ such that

$$f(a, b) \leq f(x, y) \leq f(c, d)$$

for all $(x, y) \in S$.

Theorem 25 For continuous function $f(x, y)$ on closed and bounded set $S \subset \mathbb{R}^2$, to find the maximum and minimum values of f :

1. Find all critical points of f contained in S
2. Evaluate f at each critical point
3. Find the maximum and minimum points of f on boundary $B(S)$

Theorem 26 Lagrange Multiplier Algorithm Assume that $f(x, y)$ is a differentiable function and $g \in C^1$. To find the maximum value and minimum value of f subject to the constraint $g(x, y) = k$, evaluate $f(x, y)$ at all points (a, b) which satisfy one of the following conditions.

1. $\nabla f(a, b) = \lambda \nabla g(a, b)$ and $g(a, b) = k$
2. $\nabla g(a, b) = (0, 0)$ and $g(a, b) = k$
3. (a, b) is an endpoint of the curve $g(x, y) = k$

Theorem 27 Lagrange Multipliers Generalization Let $f(\vec{x})$ be a function of n variables with r constraints of the form

$$g_1(\vec{x}) = 0, g_2(\vec{x}) = 0, \dots, g_r(\vec{x}) = 0$$

To find all possible maximum and minimum points of f subject to the constraints, find all points \vec{a} such that

$$\nabla f(\vec{a}) = \lambda_1 \nabla g_1(\vec{a}) + \dots + \lambda_r \nabla g_r(\vec{a})$$

and $g_i(\vec{a}) = 0$ for all $1 \leq i \leq r$.

Definition 31 Vector-Valued Function A function whose domain is a subset of \mathbb{R}^n and whose codomain is \mathbb{R}^m is called a vector-valued function.

Definition 32 Mapping A vector-valued function whose domain is a subset of \mathbb{R}^n and whose codomain is a subset of \mathbb{R}^n is called a mapping.

Definition 33 Derivative Matrix The derivative matrix of a mapping defined by

$$F(x, y) = (f(x, y), g(x, y))$$

is denoted DF and is defined by

$$DF = \begin{bmatrix} \frac{\partial f}{\partial x} & \frac{\partial f}{\partial y} \\ \frac{\partial g}{\partial x} & \frac{\partial g}{\partial y} \end{bmatrix}$$

Definition 34 General Mapping and Linear Approximation A mapping $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is defined by $\vec{u} = F(\vec{x}) = (f_1(\vec{x}), \dots, f_m(\vec{x}))$ for $\vec{x} \in \mathbb{R}^n$. If $F \in C^1$, the $m \times n$ derivative matrix of F is

$$DF(\vec{x}) = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \cdots & \frac{\partial f_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1} & \cdots & \frac{\partial f_m}{\partial x_n} \end{bmatrix}$$

and the linear approximation of F at \vec{a} is

$$F(\vec{x}) \approx F(\vec{a}) + DF(\vec{a})\Delta\vec{x}$$

where

$$\Delta\vec{u} = \begin{bmatrix} \Delta u_1 \\ \vdots \\ \Delta u_m \end{bmatrix} \in \mathbb{R}^m, \Delta\vec{x} = \begin{bmatrix} \Delta x_1 \\ \vdots \\ \Delta x_n \end{bmatrix} \in \mathbb{R}^n$$

Theorem 28 Chain Rule in Matrix Form Let F and G be mappings from \mathbb{R}^2 to \mathbb{R}^2 . If $G \in C^1$ at (x, y) and $F \in C^1$ at $(u, v) = G(x, y)$, then the composite mapping $F \circ G$ is C^1 at (x, y) and

$$D(F \circ G)(x, y) = DF(u, v)DG(x, y)$$

Definition 35 Invertible Mapping and Inverse Mapping Let F be a mapping from a set D_{xy} to a set D_{uv} . If there exists a mapping F^{-1} , called the inverse of F which maps D_{uv} onto D_{xy} such that

$$(x, y) = F^{-1}(u, v) \text{ iff } (u, v) = F(x, y)$$

then F is invertible on D_{xy} .

Theorem 29 One-to-One Implies Invertible Let F be a mapping from a set D_{xy} onto a set D_{uv} . If F is one-to-one on D_{xy} , then F is invertible on D_{xy} .

Theorem 30 Inverse of the Derivative Matrix Let F be a mapping from a set D_{xy} onto a set D_{uv} . If $F \in C^1$ at $\vec{x} \in D_{xy}$, and there exists an inverse mapping F^{-1} of F which has continuous partial derivatives at $\vec{u} = F(\vec{x}) \in D_{uv}$, then

$$DF^{-1}(\vec{u})DF(\vec{x}) = I$$

Definition 36 The Jacobian The Jacobian of a mapping

$$(u, v) = F(x, y) = (u(x, y), v(x, y))$$

is denoted $\frac{\partial(u, v)}{\partial(x, y)}$ and is defined by

$$\frac{\partial(u, v)}{\partial(x, y)} = \det[DF(x, y)] = \det \begin{bmatrix} \frac{\partial u}{\partial x} & \frac{\partial u}{\partial y} \\ \frac{\partial v}{\partial x} & \frac{\partial v}{\partial y} \end{bmatrix}$$

Corollary 30.1 Consider a mapping defined by

$$(u, v) = F(x, y) = (u(x, y), v(x, y))$$

which maps a subset D_{xy} onto a subset D_{uv} . Suppose $f, g \in C^1$ on D_{xy} . If F has an inverse mapping F^{-1} , with continuous partial derivatives on D_{uv} , then the Jacobian of F is non-zero:

$$\frac{\partial(u, v)}{\partial(x, y)} \neq 0 \text{ on } D_{xy}$$

Corollary 30.2 Inverse Property of the Jacobian Consider a mapping F which maps D_{xy} onto D_{uv} . If $F \in C^1$ at $\vec{x} \in D_{xy}$, and there exists an inverse mapping F^{-1} of F which is C^1 at $\vec{u} = F(\vec{x}) \in D_{uv}$, then

$$\frac{\partial(x, y)}{\partial(u, v)} = \frac{1}{\frac{\partial(u, v)}{\partial(x, y)}}$$

Theorem 31 Inverse Mapping Theorem If a mapping $(u, v) = F(x, y)$ has continuous partial derivatives in some neighbourhood of (a, b) and $\frac{\partial(u, v)}{\partial(x, y)} \neq 0$ at (a, b) , then there is a neighbourhood of (a, b) in which F has an inverse mapping $(x, y) = F^{-1}(u, v)$ which is C^1 .

Definition 37 General Jacobian Form For a mapping defined by

$$\vec{u} = F(\vec{x}) = (f_1(\vec{x}), \dots, f_n(\vec{x}))$$

where $\vec{u} = (u_1, \dots, u_n)$ and $\vec{x} = (x_1, \dots, x_n)$, the Jacobian of F is

$$\frac{\partial(u_1, \dots, u_n)}{\partial(x_1, \dots, x_n)} = \det[DF(\vec{x})] = \det \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \cdots & \frac{\partial f_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_n}{\partial x_1} & \cdots & \frac{\partial f_n}{\partial x_n} \end{bmatrix}$$

Definition 38 Integrable Function Let $D \subset \mathbb{R}^2$ be closed and bounded. Let P be a partition of D , and let $|\Delta P|$ denote the length of the longest side of all rectangles in the partition P . A function $f(x, y)$ which is bounded on D is integrable on D if all Riemann sums approach the same value as $|\Delta P| \rightarrow 0$.

Definition 39 Double Integral If $f(x, y)$ is integrable on a closed bounded set D , then we define the double integral of f on D as

$$\iint_D f(x, y) dA = \lim_{\Delta P \rightarrow 0} \sum_{i=1}^n f(x_i, y_i) \Delta A_i$$

Theorem 32 Double Integral Linearity If $D \subset \mathbb{R}^2$ is a closed and bounded set and f and g are two integrable functions on D , then for any constant c :

$$\iint_D (f + cg) dA = \iint_D f dA + c \iint_D g dA$$

Theorem 33 Double Integral Inequality If $f(x, y)$ is integrable on a closed bounded set D such that $f(x, y) \leq g(x, y)$ for all $(x, y) \in D$,

$$\iint_D f(x, y) \leq \iint_D g(x, y)$$

Theorem 34 Double Integral Absolute Value Inequality If $D \subset \mathbb{R}^2$ is a closed and bounded set and f is an integrable function on D , then

$$\left| \iint_D f dA \right| \leq \iint_D |f| dA$$

Theorem 35 Double Integral Decomposition Let $D \subset \mathbb{R}^2$ be a closed and bounded set and f be an integrable on D . If D is decomposed into two closed and bounded subsets D_1 and D_2 by a piecewise smooth curve C , then

$$\iint_D f dA = \iint_{D_1} f dA + \iint_{D_2} f dA$$

Theorem 36 Iterated Double Integrals Let $D \subset \mathbb{R}^2$ be defined by

$$y_l(x) \leq y \leq y_u(x) \text{ and } x_l \leq x \leq x_u$$

where $y_l(x)$ and $y_u(x)$ are continuous for $x_l \leq x \leq x_u$. If $f(x, y)$ is continuous on D , then

$$\iint_D f(x, y) dA = \int_{x_l}^{x_u} \int_{y_l(x)}^{y_u(x)} f(x, y) dy dx$$

Theorem 37 Let each of D_{uv} and D_{xy} be a closed bounded set whose boundary is a piecewise-smooth closed curve. Let

$$(x, y) = F(u, v) = (f(u, v), g(u, v))$$

be a one-to-one mapping of D_{uv} onto D_{xy} with $f, g \in C^1$ and $\frac{\partial(x, y)}{\partial(u, v)} \neq 0$ except for possibly on a finite collection of piece-wise smooth curves in D_{uv} . If $G(x, y)$ is continuous on D_{xy} , then

$$\iint_{D_{xy}} G(x, y) dx dy = \iint_{D_{uv}} G(f(u, v), g(u, v)) \left| \frac{\partial(x, y)}{\partial(u, v)} \right| du dv$$

The double integral theorems can be extended to triple integrals. Due to repetition, they are omitted.