ANALYSIS THEOREMS

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1. Set Theory

Here are some basic theorems from introductory analysis.

Theorem 1.1 (Cantor-Bernstein). If $f: X \to Y$ and $g: Y \to X$ are injections, |X| = |Y|.

Proof. We will construct a function $h: X \to Y$ that is a bijection. First, define $C \subset X$ as $\bigcup_{n \in \mathbb{N} \cup 0} (g \circ f)^n (A - g(B))$. Then we define

$$h(x) = \begin{cases} f & x \in C \\ g^{-1} & x \notin C \end{cases}$$

We can now check this is a bijection. For injectivity, suppose h(a) = h(b), we'd like to show a = b. As f, g^{-1} are injective, it suffices to suppose $a \in C, b \notin C$ and find a contradiction. Then $g^{-1}(b) = h(b) = h(a) = f(a) = f \circ (g \circ f)^n(c)$, but by applying g to each side of this equation we have $b \in C$, a contradiction.

For surjectivity, let $y \in Y$. Suppose $g(y) \notin C$. Then h(g(y)) = y. Now suppose $g(y) \in C$. Then $g(y) = (g \circ f)^n(c)$ with $c \notin g(B)$, which implies $n \geq 1$. Then by injectivity of g we get $y = f((g \circ f)^{n-1}(c)) = h((g \circ f)^{n-1}(c))$.

2. Inequalities

Theorem 2.1 (Cauchy-Schwarz Inequality). In an inner product space, $|(x,y)| \le ||x|| ||y||$ with equality holding iff x and y are linearly dependent.

Proof. After multiplying x by an element of S^1 , we may assume (x, y) is real. Consider $z = y - \frac{(x,y)}{\|x\|^2}x$, the projection of y onto the orthogonal complement of x. Indeed x, z are orthogonal as $(z, x) = (y, x) - \frac{(x,y)}{\|x\|^2}(x, x) = 0$. Then we have:

$$0 \le (z, z) = (z, y) - \frac{(x, y)}{\|x\|^2} (z, x) = (z, y) = (y, y) - \frac{(x, y)}{\|x\|^2} (x, y)$$

which after rearrangement is what we want. Note that the equality above happens iff z = 0 which happens iff x and y are linearly dependent.

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Theorem 2.2 (AM-GM Inequality). Let x_1, \ldots, x_n be non-negative reals. Then $\prod_{i=1}^{n} x_i \leq \left(\frac{\sum_{i=1}^{n} x_i}{n}\right)^n$ with equality holding iff $x_1 = \cdots = x_n$.

Proof. We suppose some of the x_i are not equal and show the strict inequality via induction. If μ denotes $\frac{\sum_{i=1}^{n} x_i}{n}$ then WLOG we may assume $x_n > \mu > x_{n-1}$. Then define $y = x_n + x_{n-1} - \mu$ and note y is non-negative. Then by induction, we have

$$y\prod_{1}^{n-2}x_{i} \leq \left(\frac{y+\sum_{1}^{n-2}x_{i}}{n-1}\right)^{n-1} = \mu^{n-1}$$

Multiplying by μ , we get

$$\mu y \prod_{1}^{n-2} x_i \le \mu^n$$

so it suffices to show $\mu y > x_n x_{n-1}$, but this is true as

$$y\mu - x_n x_{n-1} = (x_n + x_{n-1} - \mu)\mu - x_n x_{n-1} = (x_n - \mu)(\mu - x_{n-1}) > 0$$

Jensen's Inequality is a powerful inequality. To prove it in its measure-theoretic form, we need the notion of a subderivative and a convex function.

Definition 2.3. Let $A \subset V$ be a convex subset of a real vector space. A function $f: A \to \mathbb{R}$ is convex if $f(tx + (1-t)y) \leq tf(x) + (1-t)f(y)$ for $t \in [0,1]$. It is **strictly convex**, if the inequality is strict for $x \neq y, t \neq 0, 1$.

The following lemma is obvious.

Lemma 2.4. *f* is convex iff its graph with the set of points lying above it is convex.

Definition 2.5. Let $f : \mathbb{R} \to \mathbb{R}$ be convex. A **subderivative** of f at p is a c such that $f(x) - f(p) \ge c(x - p)$. The set of subderivatives of f at p is called the **subdifferential**.

Lemma 2.6. If f is convex, the subdifferential at p is [a,b] where $a = \lim_{x \to p^-} \frac{f(x) - f(p)}{x - p}$, $b = \lim_{x \to p^+} \frac{f(x) - f(p)}{x - p}$. Moreover, these limits exist and $a \le b$. If f is strictly convex, then f(x) - f(y) > c(x - y) when $x \ne y$ if c is a derivative.

Proof. WLOG, f(p) = p = 0. Setting y = 0 in the definition of convex, we get $f(tx) \leq tf(x)$, so $\frac{f(x)}{x}$ is increasing for all x > 0, and x < 0. Now by convexity again, $2f(0) \leq f(\epsilon) + f(-\epsilon)$ so $\frac{f(-\epsilon)}{-\epsilon} \geq \frac{f(\epsilon)}{\epsilon}$. Thus the limits a, b exist, and are finite, and it is clear that [a, b] is the subdifferential. Running through the proof for strictly convex f shows $\frac{f(x)}{x}$ is strictly increasing, so that the strict inequality holds. \square

Lemma 2.7. If f is C^2 , with $f'' \ge 0$, it is convex. If f'' > 0, it is strictly convex.

Proof. By Taylor's Theorem, one computes

$$tf(x) + (1-t)f(y) - f(tx + (1-t)y) = t(f(x) - f(tx + (1-t)y) + (1-t)(f(y) - f(tx + (1-t)y))$$
$$= tf''(c)(1-t)^{2}(y-x)^{2}/2 + (1-t)f''(c')t^{2}(y-x)^{2}/2$$

, which satisfies the correct inequalities by assumption.

Theorem 2.8 (Jensen's Inequality). If (Ω, μ) is a probability measure, $f : \mathbb{R} \to \mathbb{R}$ a convex function, g a μ -integrable function, then $f(\int_{\Omega} g d\mu) \leq \int_{\Omega} f \circ g d\mu$. If f is strictly convex, equality holds iff g takes constant value on a set of measure 1.

Proof. Define $x_0 = \int_{\Omega} g d\mu$. By Lemma 2.6 for c, there is a, b so $ax + b \ge f(x)$, $ax_0 + b = f(x_0)$. Then $f(\int_{\Omega} g d\mu) = f(x_0) = ax_0 + b = \int_{\Omega} (ag + b) d\mu \le \int_{\Omega} f \circ g d\mu$. Equality then holds iff $ag + b \ne f \circ g$ on a set of measure 0. If f is strictly convex, by Lemma 2.6, this holds iff g takes constant value on a set of measure 1.

Now that we have this inequality, we can prove many inequalities more quickly, especially exploiting the convexity/concavity of functions like log.

Definition 2.9. The weighted power mean with exponent p is the function $M_p(x_1, \ldots, x_n) = (\sum w_i x_i^p)^{1/p}$ where x_i are positive reals, and w_i are weights summing to 1.

In particular, M_{∞} is the maximum, M_2 the square mean, M_1 the arithmetic mean, M_0 the geometric mean, M_{-1} the harmonic mean, and $M_{-\infty}$ the minimum. The only one which isn't so obvious is M_0 , but to see this, by L'Hopital's Rule and continuity of the exponential function,

$$\lim_{p \to 0} \left(\sum w_i x_i^p \right)^{\frac{1}{p}} = \lim_{p \to 0} e^{\log \left(\sum w_i x_i^p \right)^{\frac{1}{p}}} = e^{\lim_{p \to 0} \frac{\log \left(\sum w_i x_i^p \right)}{p}} = e^{\lim_{p \to 0} \frac{\sum w_i x_i^p \log(x_i)}{\left(\sum w_i x_i^p \right)}}$$

$$= e^{\sum w_i \log(x_i)} = \prod x_i^{w_i}$$

The following generalizes Theorem 2.2.

Theorem 2.10 (Power Mean Inequality). Let x_i be positive reals, w_i weights. Then $p < q \implies M_p \leq M_q$, with equality holding iff the x_i with positive weights are equal.

Proof. First we will prove the inequality for the cases p = 0, q = 0. By Jensen's inequality using concavity of log, $\log \prod x_i^{w_i} = \sum \frac{w_i}{q} \log x_i^q \le \frac{\log(\sum w_i x_i^q)}{q}$ for p > 0 and q = 0 case is similar. Now it suffices to prove the inequality when pq > 0, and note that the p > 0 and p < 0 cases are equivalent since $\left(\sum w_i x_i^{-p}\right)^{\frac{1}{-p}} = \frac{1}{\left(\sum w_i \left(\frac{1}{x_i}\right)^p\right)^{\frac{1}{p}}}$.

Now note $x^{\frac{p}{q}}$ is concave, so $(\sum w_i x_i^p)^{\frac{1}{p}} = (\sum w_i (x_i^q)^{\frac{p}{q}})^{\frac{1}{p}} \leq (\sum w_i x_i^q)^{\frac{p}{q}} = (\sum w_i x_i^q)^{\frac{1}{q}}$.

We can also quickly get the Hölder inequality.

Lemma 2.11. $x, y > 0 \implies xy \le \frac{x^p}{p} + \frac{x^q}{q} \text{ with } \frac{1}{p} + \frac{1}{q} = 1.$

Proof. We give two proofs. Jensen's inequality gives $\log(xy) = \frac{\log(x^p)}{p} + \frac{\log(y^q)}{q} \le \log(\frac{x^p}{p} + \frac{y^q}{q})$. Alternatively, it is equivalent to prove $x^{\frac{1}{p}}y^{\frac{1}{q}} \le \frac{x}{p} + \frac{y}{q}$, which is homogeneous so WMA y = 1, in which case, we can optimize x to get the inequality. \square

Theorem 2.12 (Hölder Inequality). Let (Ω, Σ, μ) be a measure space, $\frac{1}{p} + \frac{1}{q} = 1$ and f, g measurable functions. Then $||fg||_1 \le ||f||_p ||g||_q$.

Proof. By Lemma 2.11,

$$||fg||_1 = \int_{\Omega} |fg| d\mu = \int_{\Omega} t|f|t^{-1}|g| d\mu \le \int_{\Omega} t^p \frac{|f|^p}{p} d\mu + \int_{\Omega} t^{-q} \frac{|g|^q}{q} d\mu = \frac{t^p}{p} ||f||_p^p + \frac{t^{-q}}{q} ||g||_q^q$$

We optimize t to get $t = \frac{\|g\|_q^{\frac{q}{p+q}}}{\|f\|_p^{\frac{p}{p+q}}}$, and plugging this in and simplifying yields the inequality.

The Hölder inequality can establish that L^p spaces are normed vector spaces. The triangle inequality is the hard part of the proof.

Lemma 2.13. $|x+y|^p \le 2^{p-1}(|x|^p + |y|^p)$ for p > 1.

Proof.
$$f(x) = x^p$$
 is convex, so $\left| \frac{x+y}{2} \right|^p \le \frac{1}{2} (|x|^p + |y|^p)$.

Theorem 2.14 (Minkowski's Inequality). $||f + g||_p \le ||f||_p + ||g||_p$.

Proof. We give two proofs, both which use Hölder inequality. By the Lemma, $||f+g||_p$ is finite if $||f||_p$, $||g||_p$ are. Then by Hölder inequality,

$$||f+g||_p^p = \int |f+g|^p \le \int |f||f+g|^{p-1} + |g||f+g|^{p-1}$$

$$\leq (\|f\|_p + \|g\|_p)\|(f+g)^{p-1}\|_{\frac{p}{p-1}} = (\|f\|_p + \|g\|_p)\|f+g\|_p^{p-1}$$

For the second proof, we claim $||f||_p = \sup_{||h||_q=1} ||fh||_1$, with which the theorem follows from the triangle inequality for L^1 . The \geq follows from the Hölder inequality, and the \geq follows from setting h to be f^{p-1} divided by its norm. Note that this proof also shows the duality between L^p and L^q (indeed they are duals as Banach spaces).

3. Topology of \mathbb{R}

Lemma 3.1. An infinite subset X of a compact Hausdorff space Y has a limit point.

Proof. If not, X is closed, and is a countable compact discrete space, which is a contradiction.

Proposition 3.2 (Baire's Category Theorem). A locally compact Hausdorff space or complete metric space X is a Baire space.

Proof. First let's treat locally compact Hausdorff. Let X_n be closed sparse subsets of X, and U_0 an open set in X. We can inductively produce $\bar{U}_i \subset U_{i-1}$ that avoid X_1, \ldots, X_i by regularity. But then $\cap_i \bar{U}_i$ is nonempty, so there is a point that avoids all U_i . Similarly in the complete metric space case, we can choose the U_i in the same way with the condition that the diameter of U_i is less than 1/n. Then again the intersection must be a single point by completeness.

Theorem 3.3. The unit cube I^n is compact.

Proof. Suppose we have a covering with no finite subcover. Then we can divide the unit cube into 2^n pieces of half the size, and at least one of these must have the same property. We can keep doing this, getting a chain of cubes contained within one another with diameter approaching 0 such that this cover has no finite subcover for any of these. But the intersection of these has diameter 0, but contains exactly 1 point by taking the limiting point of a sequence of points in each of the cubes in the chain. Taking an open set surrounding this point, we get a contradiction as infinitely many cubes must be in the open set.

Corollary 3.4 (Heine-Borel). A subset of \mathbb{R}^n is compact iff it is closed and bounded.

Proof. If a subset is compact, it is closed as \mathbb{R}^n is Hausdorff, and bounded by looking at the cover given by open balls around the origin. Conversely if a subset is closed and bounded, it is a closed subset of a large unit cube, hence is compact by Theorem 3.3.

Corollary 3.5 (Bolzano-Weierstrass). A subset of \mathbb{R}^n is compact iff any sequence has a convergent subsequence.

Proof. 'Suppose any sequence has a convergent subsequence. Then the subset must be bounded or else we could take an increasingly large sequence. It must also be closed or else we could take points in decreasingly small neighborhoods of a limit point. The converse follows from Lemma 3.1.

Lemma 3.6 (Minimum-Maximum Theorem). A map $f: S \to \mathbb{R}$ from compact S has a min and max.

6

Proof. The image is compact, hence closed and bounded, so attains its supremum and infimum. \Box

Theorem 3.7. All norms on finite dimensional \mathbb{R} -vector spaces are equivalent.

Proof. We will show every norm is equivalent to $\|\cdot\|_{\infty}$. To do this, let $\|\cdot\|$ be any norm, and we will show it is continuous with respect to $\|\cdot\|_{\infty}$. In particular, if $M = \max_i \|\delta_i\|$ we have $\|x\| \leq \sum_1^n \|x_i\delta_i\| \leq M\sum_1^n |x_i| \leq Mn\|x\|_{\infty}$, so we have $\|x\| - \|a\| \leq \|x - a\| \leq Mn\|x - a\|_{\infty}$ so indeed it is continuous.

The set $X = \{x | ||x||_{\infty} = 1\}$ is compact, so its image from $||\cdot||$ by Lemma 3.6 has a minimum m and a maximum M. Thus m and M give the bounds we need for equivalence of norms.

Lemma 3.8. There is no retraction from $D^n \to S^n$.

Proof. Look at H^{n-1} .

Proposition 3.9 (Intermediate Value Theorem). A map from D^n to itself that is the identity on the boundary is surjective.

Proof. If not, we can deformation retract away from the missing point to the boundary, to yield a retraction of D^n onto S^{n-1} , contradicting Lemma 3.8. In the case n=1 this is looking at the connected component.

Proposition 3.10 (Brouwer Fixed Point Theorem). A map $f: D^n \to D^n$ has a fixed point.

Proof. If not, we can consider the line going from f(x) to x and produce a retraction g that is the intersection (on the x side) of this line with the boundary, contradicting Lemma 3.8.

Theorem 3.11. For any embedding $i: D^k \to S^n$, $S^n - i(D^k)$ has trivial \tilde{H}_* .

Proof. We induct on k, for k=0 we just have $S^n-i(D_0)=\mathbb{R}^n$. For the induction step, we replace D^k with I^k . Now we split I^k into two halves $I^{k-1}\times [0,\frac{1}{2}], I^{k-1}\times [\frac{1}{2},1]$ and by Mayer-Vietoris and induction we have $\tilde{H}_*(S^n-I^k)\hookrightarrow \tilde{H}_*(I^{k-1}\times [0,\frac{1}{2}])\oplus \tilde{H}_*(I^{k-1}\times [\frac{1}{2},1])$ is an isomorphism. If there were a nontrivial (reduced) cycle α in $H_*(S^n-I^k)$ it would land nontrivially in one of the two summands on the right, and we could repeatedly cut these up into smaller intervals, in the limit landing in $H_*(S^n-I^{k-1})$ in which we know it would be a boundary of a chain β . But then as β is compact and covered by our cuts, it is in one of the cuts, a contradiction. \square

The next theorem has the Jordan Curve Theorem as a special case:

Theorem 3.12. For any embedding $i: S^k \to S^n, n > k$ we have $\tilde{H}_i(S^n - i(S^k)) = \mathbb{Z}^{\delta_{i,n-k-1}}$.

Proof. We induct on k, for k = 0 we have $S^n - i(S^0) \simeq S^{n-1}$. Now to induct we split S^k into to disks and use Mayer-Vietoris and Theorem 3.11 to give $H_{m-1}(S^n - i(S^k)) \cong H_m(S^n - i(S^{k-1}))$.

Theorem 3.13 (Invariance of Domain). An injective map $\mathbb{R}^n \to \mathbb{R}^n$ is open.

Proof. By Theorem 3.12, the boundary of an embedding $D^n \to \mathbb{R}^n$ separates \mathbb{R}^n into two components, and since the boundary is closed, the interior of D^n is an connected component of this separation that is open in \mathbb{R}^n .

Lemma 3.14 (Lebesgue Number Lemma). Any open cover of a compact metric space X has a δ (Lebesgue number) so that every δ ball is in some open set.

Proof. Take a finite subcover U_1, \ldots, U_n , and consider the map $f: X \to \mathbb{R}$, $f(x) = \sum d(x, X - U_i)$. Now this function attains a minimum δ , but δ cannot be 0 as the U_i cover X. Then δ is a Lebesgue number.

Proposition 3.15. A continuous map $f: S \to T$ where S, T are metric spaces, and S is compact is uniformly continuous.

Proof. Take the preimages of ϵ balls around each point in T to make a cover of S. The Lebesgue number of this cover is a uniform δ for this ϵ .

Theorem 3.16 (Uniform Limit Theorem). A uniform limit of continuous functions to a metric space is continuous.

Proof. Let the sequence be f_n and the limit be f. We need that for any $\epsilon > 0$ and any $x \in X$ there is a neighborhood $x \in U_{\epsilon}$ such that $d(f(x), f(y)) < \epsilon$ when $x, y \in U_{\epsilon}$. We can choose large enough n so we have $\forall x, d(f(x), f_n(x)) < \epsilon$, and a neighborhood U satisfying ϵ for f_n Then we have in U:

$$d(f(x), f(y)) \le d(f(x), f_n(x)) + d(f_n(x), f_n(y)) + d(f_n(y), f(y)) < \epsilon + \epsilon + \epsilon = 3\epsilon$$
 so we are done.

Theorem 3.17. If a sequence of continuous functions f_n from a compact space X to a metric space Y converge to a function f, they uniformly converge to that function, which is continuous by Theorem 3.16.

Proof. For any $\epsilon > 0$, there is a neighborhood around each point $x \in X$, U_x such that f(y) is within $\epsilon/2$ of f(x) for any $y \in U_x$. Then make a finite cover of these U_x and take the maximum N for each of the corresponding points such that f_N is within $\frac{\epsilon}{2}$ for these points. Then by the triangle inequality every point simultaneously satisfies $|f(x) - f_N(x)| \le \epsilon$ for large enough N.

Theorem 3.18 (Ascoli). If $f_n: X \to Y$ form a uniformly equicontinuous sequence of maps between compact metric spaces X and Y, there is a uniformly convergent subsequence.

8

Proof. X has a countable dense subset x_n , so starting with x_1 we can look at $f_i(x_1)$, which has a limit point by Lemma 3.1. This gives a subsequence which converges at x_1 . We can then inductively produce more subsequences for each x_n and take a diagonal sequence g_n . Every compact metric space is complete by Lemma 3.1, so it suffices to show that g_n is uniformly Cauchy. To do so, fix $\epsilon > 0$, and cover X with finitely many neighborhoods around the x_i so each f_n varies at most $\frac{\epsilon}{3}$ in the neighborhood. Each neighborhood contains an x_i so choose the maximum of the Ns so for $j \geq N$ the $g_j(x_i)$ differ by at most $\frac{\epsilon}{3}$. Then for any $x \in X$, j,k > N we have $d(g_j(x), g_k(x)) \leq d(g_j(x), g_j(x_i)) + d(g_j(x_i), g_k(x_i)) + d(g_k(x_i), g_k(x)) < \epsilon$.

4. Derivatives

Let \mathbb{F} from now on denote \mathbb{C} or \mathbb{R} .

Proposition 4.1. A function $f : \mathbb{R}^n \to \mathbb{R}$ that has a local extremum at a point $a \in U$ and is differentiable there has a critical point.

Proof. It suffices to consider n=1, as this implies each of the $\partial_i f(a)$ in the general case is 0. In this case, note that if |f'(a)| > 0 then for small enough ϵ we have $\frac{|\epsilon|}{|h|} < |f'(a)|$ so the function decreases on one side and increases on the other.

Proposition 4.2 (Mean Value Theorem). If $f, g : [a, b] \to \mathbb{R}$ are continuous, and differentiable on the interior, then there is a c so that

$$(g(b) - g(a))f'(c) = (f(b) - f(a))g'(c)$$

Proof. Consider the function h = (g(b) - g(a))f(x) - (f(b) - f(a))g(x). h(a) = g(b)f(a) - f(b)g(a) = h(b), so either h is constant for which any point works, or it attains some maximum/minimum on the interior of the interval, which by Proposition 4.1 yields a point c with h'(c) = 0.

Note that the Mean Value Theorem is most often used in the case where g(x) = x. The chain rule is the theorem that in the category Diff, d is an endofunctor that takes a manifold to its tangent space, and a map f to a map df (the pushforward, total derivative, or derivative) on tangent spaces. In the case of a map between open sets of \mathbb{F}^n , the tangent space is canonically identified with $\mathbb{F}^n \times \mathbb{F}^n$. Here is the classical statement:

Theorem 4.3 (Chain Rule). Suppose we have $U \subset \mathbb{F}^l, V \subset \mathbb{F}^m, W \subset \mathbb{F}^n$, and there are maps $f: U \to V, g: V \to W$ such that f is differentiable at a, and g is differentiable at a. Then $a \circ f$ is differentiable and $a \circ f$ is dif

Proof. We define k = f(a+h) - f(a). Then:

$$(g \circ f)(a+h) - (g \circ f)(a) = g(f(a+h)) - g(f(a)) = g(f(a)+k) - g(f(a))$$

$$= g'(f(a))k + \epsilon_g(k) = g'(f(a))(f(a+h) - f(a)) + \epsilon_g(k) = g'(f(a))(f'(a)h + \epsilon_f(h)) + \epsilon_g(k)$$
$$= g'(f(a))f'(a)h + \epsilon_g(h) = g'(f(a))(f'(a)h + \epsilon_f(h)) + \epsilon_f(h) = g'(h) + \epsilon_f(h) +$$

where $\epsilon = g'(f(a))\epsilon_f(h) + \epsilon_g(k)$ so it suffices to show $\frac{\|\epsilon\|}{\|h\|} \to 0$ as $h \to 0$. With $\|\cdot\|_o$ denoting the operator norm we have:

$$\frac{\|\epsilon\|}{\|h\|} \le \frac{\|g'(f(a))\|_o \|e_f(h)\|}{\|h\|} + \frac{\|e_g(k)\|}{\|k\|} \frac{\|f(a+h) - f(a)\|}{\|h\|}
= \frac{\|g'(f(a))\|_o \|e_f(h)\|}{\|h\|} + \frac{\|e_g(k)\|}{\|k\|} \left(\frac{\|f'(a)\|_o \|h\|}{\|h\|} + \frac{\|\epsilon_f(h)\|}{\|h\|}\right)$$

which tends to 0 by hypothesis.

Corollary 4.4 (Product & Quotient Rules).

$$\partial_i(f(x)g(x)) = \partial_i f(x)g(x) + \partial_i g(x)f(x)$$
$$\partial_i \frac{f(x)}{g(x)} = \frac{\partial_i f(x)g(x) - \partial_i g(x)f(x)}{g(x)^2}$$

Proof. Use the chain rule on the composite $x \mapsto (f(x), g(x)) \mapsto f(x)g(x)$ and $x \mapsto (f(x), g(x)) \mapsto \frac{f(x)}{g(x)}$.

Corollary 4.5. If the inverse of f is differentiable, $(f^{-1})'(f(x)) = f'(x)^{-1}$.

Proof. Apply the chain rule to $f^{-1} \circ f$.

Corollary 4.6 (Mean Value Theorem Many Variables). If $f : \mathbb{R}^n \to \mathbb{R}$ is continuous at a, b and differentiable on a neighborhood which contains the line segment strictly between a and b, then there is a c on this line segment satisfying f(b) - f(a) = f'(c)(b-a).

Proof. Consider the function $g:[0,1] \to U$ going to the straight line between a and b. By Proposition 4.2 we have $(f \circ g)'(c) = f(b) - f(a)$, and we can use the chain rule to get $(f \circ g)'(c) = f'(g(c))g'(c) = f'(g(c))(b-a)$.

Corollary 4.7 (Mean Value Inequality). If $f: \mathbb{R}^m \to \mathbb{R}^n$ is continuous at a, b and differentiable on a neighborhood which contains the line segment strictly between a and b, then there is a c on the line segment such that $||f(b)-f(a)||_2 \le ||f'(c)||_0 ||b-a||_2$.

Proof. Let u be the unit vector in the direction of f(b) - f(a). Then let $U(x) = u \cdot x$, so $U \circ f$ is a real valued function to which we can apply Corollary 4.6. We get $(U \circ f)(b) - (U \circ f)(a) = (U \circ f)'(c)(b-a) = U'(f(c))f'(c)(b-a) = u \cdot (f'(c)(b-a))$. Then we have via Cauchy-Schwarz inequality,

$$||f(b) - f(a)||_2 = |(U \circ f)(b) - (U \circ f)(a)| = |u \cdot (f'(c)(b - a))|$$

$$< ||u||_2 ||f'(c)(b - a)|| < ||f'(c)||_0 ||b - a||_2$$

Corollary 4.8. If $f: \mathbb{R}^m \to \mathbb{R}^n$ is differentiable in a convex bounded open set U, and $||f'(x)||_o$ is bounded on U, then f is uniformly continuous.

Proof. Corollary 4.7 gives
$$||f(x) - f(y)||_2 \le M||x - y||$$
.

Corollary 4.9 (L'Hôpital's Rule). If $f, g : \mathbb{R}^n \to \mathbb{R}$ are continuous, differentiable in a deleted neighborhood of 0, f(0) = g(0) = 0, and $\lim_{x\to 0} g'(x) \neq 0$, $\lim_{x\to 0} f'(x)$ exist, then

$$\lim_{x \to 0} \frac{f(x)}{g(x)} = \lim_{x \to 0} \frac{f'(x)}{g'(x)}$$

Proof. Use Proposition 4.2 on a small interval $(-\delta, \delta)$ around the origin, to get $\frac{f(x)}{g(x)} = \frac{f(x) - f(0)}{g(x) - g(0)} = \frac{f'(x_0)}{g'(x_0)}$ for some x_0 in the interval. Letting $\delta \to 0$ we are done.

Theorem 4.10. If $f: \mathbb{F}^m \to \mathbb{F}^n$ has all partial derivatives which are continuous near a point a, then f'(a) exists.

Proof. It suffices to prove this when n=1 as being differentiable is equivalent to being differentiable in each component. Now fix an h in a small enough ball. We have by Proposition 4.2:

$$f(a_1 + h_1, \dots, a_i + h_i, a_{i+1}, \dots, a_n) - f(a_1 + h_1, \dots, a_{i-1} + h_{i-1}, a_i, \dots, a_n)$$

= $h_i \partial_i f(a_1 + h_1, \dots, a_{i-1} + h_i, \alpha_i, a_{i+1}, \dots, a_n)$

where $\alpha_i \in (a_i, a_i + h_i)$. Now we have

$$f(a+h) - f(a) = \sum_{i} h_{i} \partial_{i} f(a_{1} + h_{1}, \dots a_{i-1} + h_{i-1}, \alpha_{i}, a_{i+1} + h_{i+1}, \dots, a_{n})$$
$$= h(\partial_{i} f(a)) + \epsilon$$

where

$$\frac{\|\epsilon\|}{\|h\|} \le \sum_{i} \|\partial_{i} f(a_{i} + h_{i}, \dots, \alpha_{i}, \dots, a_{n}) - \partial_{i} f(a)\|$$

which tends to 0 as $h \to 0$ by continuity of the partial derivatives.

Proposition 4.11. If $f : \mathbb{F}^2 \to \mathbb{F}$ is continuous near a point a, has partials $\partial_1 f, \partial_2 f, \partial_1 \partial_2 f, \partial_2 \partial_1 f$ near a, and the mixed partials $\partial_1 \partial_2 f, \partial_2 \partial_1 f$ are continuous near a, then they are equal at a.

Proof. By Proposition 4.2, we have

$$\partial_1 \partial_2 f(x'', y'') = \partial_2 f(x + h, y'') - \partial_2 f(x, y'')$$

$$= f(x + h, y + k) - f(x + h, y) - f(x, y + k) + f(x, y)$$

$$= f(x + h, y + k) - f(x, y + k) - f(x + h, y) + f(x, y)$$

$$= \partial_1 f(x', y + k) - \partial_1 f(x', y) = \partial_2 \partial_1 f(x', y')$$

so letting $k, h \to 0$ by continuity we are done.

Proposition 4.12 (One-variable Taylor Expansion). If $f : \mathbb{R} \to \mathbb{R}$ has continuous partial derivatives up to $k + 1^{th}$ order near a, then for small enough h we have

$$f(a+h) = \sum_{i=0}^{k} \frac{1}{i!} f^{(i)}(a)h^{i} + \frac{1}{(k+1)!} f^{(k+1)}(\alpha)h^{k+1}$$

with $\alpha \in (a, a + h)$.

Proof. Let R(x) be $f(a+x) - \sum_{0}^{k} \frac{1}{i!} f^{(i)}(a) x^{i} - \frac{1}{(k+1)!} c x^{h+1}$ such that c is chosen so that R(h) = 0. We want to show $c = f^{(k+1)}(\alpha)$ as above. To do this, use Proposition 4.2 many times, first on the fact that R(0) = R(h) = 0 to get a a_1 where $R'(a_1) = 0$, and then repeating this on $R'(0) = R'(a_1) = 0$. Then finally we will have a a_{k+1} with $R^{(k+1)}(a_{k+1}) = 0$, at which point taking the $k+1^{th}$ derivative yields $0 = R^{(k+1)}(a_k) = f^{(k+1)}(a + a_{k+1}) - c$, so we are done.

Corollary 4.13 (Many-variable Taylor Expansion). If $f : \mathbb{R}^n \to \mathbb{R}$ has continuous partial derivatives up to $k + 1^{th}$ order near a, for small enough h we have

$$f(a+h) = \sum_{i_1, \dots, i_l=1}^k \sum_{j_1, \dots, j_{l+1}=1}^n \partial_{j_1} \dots \partial_{j_i} f(a) \prod_{l=1}^i h_l + \sum_{j_1, \dots, j_{k+1}=1}^n \partial_{j_1} \dots \partial_{j_{k+1}} f(\alpha) \prod_{l=1}^{k+1} h_l$$

with α in the line segment between a and a + h.

Proof. Apply the chain rule to extend the one-variable case by composing f with the line between a and a + h.

5. Series

Lemma 5.1 (Raabe's Test). For the series $\sum_{0}^{\infty} c_n$, $c_i \in \mathbb{R}$ if $R_n = n(1 - \frac{c_{n+1}}{c_n})$ is larger than R > 1 for sufficiently large n, the series is absolutely convergent.

Proof. WLOG $c_n \geq 0$. Then note that $R_n - 1 = \frac{1}{c_n}(c_n(n-1) - c_{n+1}n)$ so $0 \leq c_n = \frac{c_n(n-1)-c_{n+1}n}{R_n-1}$. By hypothesis $R_n - 1$ stays away from 0 for large n so it suffices to show $c_n(n-1)-c_{n+1}n$, which is positive for large n, converges. But $\sum_{1}^{m} c_n(n-1)-c_{n+1}n = -c_{m+1}m$ is monotonically increasing for large m and bounded above by 0 so converges to a limit.

Proposition 5.2. Every power series $\sum_{0}^{\infty} c_n z^n$, $c_n \in \mathbb{C}$ has a radius of convergence R given by Hadamard's formula $\frac{1}{R} = \limsup |c_n|^{\frac{1}{n}}$

Proof. For any r < R and large n we have $|c_n|^{\frac{1}{n}} < \frac{1}{r}$ so if |z| < r < R we have $\sum_{n=1}^{\infty} |c_n| z^n < \sum_{n=1}^{\infty} \frac{z^n}{r^n}$ so we have absolute convergence. Similarly if |z| > r > R there are infinitely many terms so that $|c_n| \ge \frac{1}{r^n}$ so the series diverges.

Proposition 5.3. Every power series $\sum_{0}^{\infty} c_n z^n$ is holomorphic in its radius of convergence, and its derivative, $\sum_{0}^{\infty} n c_n z^{n-1}$ has the same radius of convergence.

Proof. First note by Proposition 5.2 since $n^{1/n} \to 1$ as $n \to \infty$ we have that $\sum_{n=0}^{\infty} nc_n z^{n-1}$ has the same radius of convergence.

Let $z_0, z_0 + h$ lie in radius r, which is less than the radius of convergence.

$$\left| \frac{\sum_{0}^{\infty} c_{j}(z_{0} + h)^{j} - \sum_{0}^{\infty} c_{j}z_{0}^{j}}{h} - \sum_{0}^{\infty} c_{j}jz_{0}^{j-1} \right|$$

$$= |\epsilon_{n}(h)| + \left| \frac{\sum_{n=1}^{\infty} c_{j}(z_{0} + h)^{j} - \sum_{n=1}^{\infty} c_{j}z_{0}^{j}}{h} \right| + |\sum_{n=1}^{\infty} c_{j}jz_{0}^{j-1}|$$

where $\epsilon_n(h)$ is the error from the n^{th} partial sum's approximation by its derivative. Since $a^j - b^j = (a - b)(a^{j-1} + a^{j-2}b + \cdots + b^{j-1})$ we have $|\frac{(z_0 + h)^j - z_0^j}{h}| \leq jr^{j-1}$. For any $\epsilon > 0$, we can choose n large enough so that $\sum_{n=1}^{\infty} jc_j r^{j-1} < \epsilon$. Afterwards we can choose h small enough so $\epsilon_n(h) < \epsilon$. Then continuing from above we get

$$\leq \epsilon + \epsilon + \epsilon = 3\epsilon$$

concluding the proof.

Proposition 5.4 (Abel's Theorem). If a power series $f(x) = \sum_{i=0}^{\infty} c_i x^i$ converges at an endpoint of its radius of convergence R, it converges uniformly and hence is continuous on [0, R].

Proof. WLOG, R = 1. Let $A_n = \sum_{i=0}^{n-1} c_i$. We have then for $x \in [0, 1]$

$$\left| \sum_{n=1}^{m} c_{i} x^{i} \right| = \left| \sum_{n=1}^{m} A_{i+1} x^{i} - \sum_{n=1}^{m} A_{i} x^{i} \right| = \left| \sum_{n=1}^{m} (f(1) - A_{i+1}) x^{i} - \sum_{n=1}^{m} (f(1) - A_{i}) x^{i} \right|$$

$$= \left| \sum_{n=1}^{m+1} (f(1) - A_{i}) x^{i-1} - \sum_{n=1}^{m} (f(1) - A_{i}) x^{i} \right|$$

$$= \left| \sum_{n=1}^{m} (f(1) - A_{i}) (x^{i-1} - x^{i}) - (f(1) - A_{i}) x^{n} + (f(1) - A_{i}) x^{m} \right|$$

now choosing N large enough so $|f(1) - A_i| < \epsilon$ for i > n and noting $x \in [0, 1]$

$$\leq \sum_{n+1}^{m} |(f(1) - A_i)|(x^{i-1} - x^i) + |(f(1) - A_i)|x^n + (f(1) - A_i)x^m|$$

$$\leq (x^n - x^m)\epsilon + x^n\epsilon + x^m\epsilon \leq 3\epsilon$$

and we are done as the partial sums are uniformly Cauchy.

Proposition 5.5. If $f(x) = \sum_{0}^{\infty} c_i z^i$ converges within radius R, and a is within this radius, f(a+x) converges and is analytic within radius of R-|a|.

Proof. Inside |z| < R - |a| the power series absolutely converges, and we would like to recenter around 0. WLOG a is real and non-negative. For 0 < x < R - |a| we have

$$\sum_{0}^{\infty} c_{i}(a+x)^{i} = \sum_{0}^{\infty} c_{i} \sum_{j=0}^{i} {i \choose j} a^{i-j} x^{j}$$

We can then split the inner sum up into separate terms, still absolutely convergent as everything is non-negative, and then we can collect like powers of x to get

$$\sum_{0}^{\infty} c_i (a+x)^i = \sum_{0}^{\infty} \left(\sum_{m}^{\infty} c_i \binom{i}{m} a^{i-m} \right) x^m$$

so f(a+z) has a power series that converges on the interval (0, R-|a|). Hence its radius of convergence is at least R-|a|.

Lemma 5.6. The composite of two \mathbb{F} -analytic functions $\mathbb{F} \to \mathbb{F}$ is analytic.

Proof. Real analytic functions are restrictions of complex analytic ones, and by the chain rule for holomorphic functions, the composite is analytic. \Box

Lemma 5.7. $(1+z)^a$ is holomorphic within radius 1 around the origin, its power series given by $\sum_{0}^{\infty} {a \choose j} z^j$. It converges absolutely and uniformly on the interval [-1,1] when a>0.

Proof. That this is analytic follows from Lemma 5.6, and that it is holomorphic of radius 1 comes from Hadamard's formula. Now when $z \in [-1,1]$, for n > a > 0 we have $n\left(1-\left|\frac{\binom{a}{n+1}}{\binom{a}{n}}z\right|\right) \geq n\left(1-\left|\frac{\binom{a}{n+1}}{\binom{a}{n}}\right|\right) = n\left(1-\frac{n-a}{n+1}\right) = \frac{n(1+\alpha)}{n+1}$ which is larger than 1 for sufficiently large n so by Lemma 5.1 and Theorem 3.17 we have the last statement.

Lemma 5.8. The function |x| can be uniformly approximated by polynomials in a bounded interval.

Proof. WLOG the interval is [-1,1]. By Lemma 5.7 $(1-t)^{1/2}$ is uniformly approximated by its Taylor expansion in this interval, and we can set $t=1-x^2$ we get $(x^2)^{1/2}=|x|$ is as well.

Theorem 5.9 (Stone-Weierstrass). If S is compact, any \mathbb{R} -subalgebra of $\mathcal{C}(S,\mathbb{R})$ that separates points is dense.

Proof. Let A be a \mathbb{R} -subalgebra of $\mathcal{C}(S,\mathbb{R})$ that separates points. If $f \in \bar{A}$, then $|f| \in \bar{A}$ as f has a bounded image so by Lemma 5.8 |f| can be approximated uniformly by polynomials in f, which are in \bar{A} .

As a consequence, $\max(f, g)$, $\min(f, g) \in \bar{A}$ whenever f, g are as they are linear combinations of f, g, |f|, |g|.

Now for any $\epsilon > 0$, $h \in \mathcal{C}(S, \mathbb{R})$ we can approximate h up to ϵ as follows: for each $a \in S$ choose $f_{a,b}$ for each b so that f(a) = h(a), f(b) = h(b). Then in a small neighborhood of b, $f_{a,b}(x) > h(x) - \epsilon$, but S is compact so finitely many such neighborhoods suffice to have $f_a = \max_{b_i} f_{a,b_i}$ be at least $h(x) - \epsilon$ everywhere. Now we can similarly choose small neighborhoods for each a so that $f_a < h(x) + \epsilon$, and once again finitely many suffice so that $\min_{a_i} f_{a_i}$ is our desired approximation. \square

There is an analogous version for $\mathcal{C}(S,\mathbb{C})$.

Corollary 5.10. If S is compact, any C^* -subalgebra of $\mathcal{C}(S,\mathbb{C})$ that separates points is dense.

Proof. Let A again be such a subalgebra. As A is closed under conjugation, it contains the real and imaginary parts of any element f. Then by Theorem 5.9 the real and imaginary parts of any function h are in \bar{A} so h is as well.

6. Endomorphisms

Lemma 6.1 (Adjugates). Let M be a free R-module. Given $f \in \operatorname{End}_R(M)$, there is an element $\operatorname{adj}(f)$ such that $f \operatorname{adj}(f) = \operatorname{adj}(f) f = 1_M$.

Proof. It suffices to prove this in the universal ring $R = \mathbb{Z}[x_{ij}]$, $1 \leq i, j \leq n$, with an endomorphism f given by the matrix (x_{ij}) . We have a natural isomorphism $j: M \cong \operatorname{Hom}(\wedge^{n-1}M, \wedge^n M)$ so that $j(x_1)(x_2 \wedge \cdots \wedge x_n) = x_1 \wedge \cdots \wedge x_n$. To get adj f, take the endomorphism corresponding to $\operatorname{Hom}(\wedge^{n-1}f, 1_{\wedge^n M})$. Now we have $j(\operatorname{adj}(f)f)$ takes $x_2 \wedge \cdots \wedge x_n \mapsto f(x_2) \wedge \cdots \wedge f(x_n) \mapsto f(x_1) \wedge \cdots \wedge f(x_n)$ which is $j(\det(f)1_M)$. Note that since $\det(f)$ is nonzero, $\operatorname{adj}(f)$ has nonzero determinant, so is injective as R is an integral domain. Now we have $\operatorname{adj}(f)f \operatorname{adj}(f) = \operatorname{adj}(f) \det(f)$ so by injectivity $f \operatorname{adj}(f) = \det(f)1_M$.

Lemma 6.2. If $Av = \lambda v$ then for any polynomial p, $p(A)v = p(\lambda)v$.

Proof. $p(A)v = p(\lambda)v$ is a linear combination of $A^nv = \lambda^nv$.

Theorem 6.3 (Cayley-Hamilton). Let M be a finitely generated free R-module, and $f \in \operatorname{End}(M)$. Then $\chi_f(f) \equiv 0$

Proof. f turns M into a R[T]-module, and we can extend scalars via $R' = R[x] \otimes_R R[T]$ and $M' = R[x] \otimes_R M$. Then $\chi_f(x)$ is the determinant of $y = x \otimes 1_M - 1 \otimes T \in R'$. adj(y) commutes with y by Lemma 6.1, and since $x \otimes 1_M$ is central it commutes with

 $1 \otimes T$ as well, but then it commutes with all of R'. Now we look at R'/(y), M'/(y)M' which substitutes x as T, and note that $\mathrm{adj}(y)$ has a well-defined action on the quotient as it commutes with R'. $M'/(y)M' \cong 1 \otimes M$ since $g(x) \otimes m = (g(x) \otimes 1)(1 \otimes m) = (1 \otimes g(T))(1 \otimes m) = 1 \otimes g(T)m$. Then since y annihilates M'/(y)M', y adj(y) does as well, but this is multiplication by $\chi_f(x) \otimes 1 = 1 \otimes \chi_f(T)$, which is the action of $\chi_f(f)$.

Corollary 6.4 (Determinant Trick). If f is an endomorphism of M, an R-module generated by n elements, and $fM \subset IM$, f satisfies $f^n + a_1 f^{n-1} + \cdots + a_n \equiv 0$ where $a_i \in I^i$.

Proof. By projectivity of free modules, it suffices to consider a free module, but then this follows from Theorem 6.3 by noting that the coefficients of $\chi_f(x)$ are of the form described.

Corollary 6.5 (Nakayama's Lemma). If M is a finitely generated R-module and IM = M, then there is an $a \equiv 1 \pmod{I}$ such that aM = 0.

Proof. Apply Corollary 6.4 to the identity map 1_M and use the fact that $1_M M \subset IM$.

Corollary 6.6 (Nakayama's Lemma). If M is a finitely generated module over a local ring R with maximal ideal m and mM = M, then M = 0.

Proof. By Corollary 6.5 aM = 0 for $a \equiv 1 \pmod{m}$ but then aM = M.

Corollary 6.7 (Nakayama's Lemma). If M is a finitely generated module over a local ring R with maximal ideal m and R and M = N + mM then M = N.

Proof. Apply Corollary 6.6 to M/N.

Corollary 6.8 (Nakayama's Lemma). If M is a finitely generated module over a local ring R with maximal ideal m and the image of m_1, \ldots, m_n generate M/mM, then m_1, \ldots, m_n generate M.

Proof. Apply Corollary 6.7 with $N = \sum_{i=1}^{n} m_i M$.

Note if the ring is not local, we can replace m by the Jacobson radical and Nakayama's Lemma still holds.

Proposition 6.9. Every endomorphism $f: V \to V$ on a finite dimensional vector space V over F has a minimal polynomial μ_f , satisfying $\mu_f(f) = 0$, $g(f) = 0 \Longrightarrow \mu_f|g$, and its roots are the eigenvalues.

Proof. Viewing V as a F[T] module, since F[T] is a PID, everything is immediate except the last part, which follows since $f(v) = \lambda v$ so by Lemma 6.2 $\mu_f(f)(v) = \mu_f(\lambda)v$ but the LHS is 0 and v is not so we are done.

Proposition 6.10. Submodules M of R^n , a finite generated free module over a PID are after a change of basis of the form $\bigoplus_{i=1}^{n} x_i r_i R$ with $x_i | x_{i+1} \in R$ and $\bigoplus_{i=1}^{n} r_i R = R^n$ (the r_i are the change of basis). This representation is unique up to units and the x_i are called the **invariant factors**.

Proof. Choose a map $f_1: R^n \to R$ where the image of M is maximized (this uses PID). Let y_1 be an element sent to a generator of the image, x_1 , which WLOG is nonzero. Now if π_i is the i^{th} projection, then $x_1|\pi_i(y_1)$ for all $x \in R^n$ by maximality of f_1 , so we can let $r_1 = \frac{y_1}{x_1}$. Now r_1 gets sent to 1 by f, so we can project orthogonal to r_1 via a section $s_1: R \to R^n$ taking $1 \mapsto r_1$. Our projection $o_1(x) = x - s_1 \circ f_1(x)$. This section gives $R^n = r_1 R \oplus o_1(R)$. The projection to $o_1(R)$ is surjective, and by removing an appropriate generator and localizing at (0), we see that our new module must be free of rank n-1. Now we apply induction to get y_2, \ldots, y_n and r_2, \ldots, r_n , and by looking at f_1 we get $x_1|x_2$. Uniqueness also follows from induction. \square

Corollary 6.11 (Smith Canonical Form). A map $f: M \to N$ between finitely generated free modules over a PID of ranks n and m has a Smith Canonical Form, i.e. is represented by a matrix of the form

$$\begin{pmatrix} x_1 & & \\ & \ddots & \\ & & x_{\min(n,m)} \end{pmatrix}$$

with $x_i|x_{i+1}$. This representation is unique up to units.

Proof. The image is a submodule of \mathbb{R}^m so this is a reformulation of Proposition 6.10.

Corollary 6.12 (Finitely Generated Modules over PIDs). A finitely generated module over a PID is of the form $R^m \oplus \bigoplus_{i=1}^n R/(d_i)$ where $d_i|d_{i+1}$. Moreover m is unique, and the d_i are unique up to units.

Proof. A finitely generated module over a PID is a quotient of a finite rank free module, which has the correct form according to Proposition 6.10.

Corollary 6.13 (Rational Canonical Form). Every endomorphism $f: V \to V$ of a finite dimensional vector space over F has a unique Rational Canonical Form, ie. is represented by a matrix of the form

$$\bigoplus_{i=1}^{n} \begin{pmatrix} 0 & 0 & 0 & \dots & -a_0 \\ 1 & 0 & 0 & \dots & -a_1 \\ 0 & 1 & 0 & \dots & -a_2 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & -a_{k_i-1} \end{pmatrix}$$

where the monic polynomials (invariant factors) $f_i(x) = \sum_{1}^{k_i} a_i x^i$ satisfy $f_i|f_{i+1}$. f_n is $\mu_f(x)$ and $\prod_i f_i$ is $\chi_f(x)$

Proof. View V as a module over F[T], and we get $V \cong \sum_{i=1}^{n} F[T]/(f_i)$ from Corollary 6.12. Then we are done by picking $1, T, \ldots, T^{k_i-1}$ as a basis.

Corollary 6.14. If A a matrix over F, the invariant factors can be computed by finding the Smith Canonical Form of xI - A.

Proof. If V is dimension n we can consider the F[T] module homomorphism $F[T]^n \to V$ mapping the generators r_i surjectively onto an F-basis v_i of V. Now the elements $y_i = Tr_j - \sum_{1}^{i} (a_{ij}r_i)$ are in the kernel but note that $\sum_{i} y_i F[T] + \sum_{i} r_i F = \sum_{i} r_i F[T] = F[T]^n$, so y_i actually generate the kernel. The y_i have the relations matrix $xI - A^t$, so by Corollary 6.11 after a change of basis it is in Smith Normal Form with invariant factors f_1, \ldots, f_n , so the kernel is of this form for an appropriate set of generators, and $V \cong \bigoplus_{1}^{n} F[T]/(f_n)$.

Note that Corollary 6.12 also can be represented as $R^m \oplus \bigoplus R/(p^i)$ where p varies over primes, and similarly Corollary 6.13 has a representation in this way.

Corollary 6.15 (Jordan Canonical Form). Every endomorphism $f: V \to V$ of a finite dimensional vector space over F has a unique Jordan Canonical Form after extending scalars to an algebraic closure, ie. is represented by a matrix of the form

$$\bigoplus_{i=1}^{n} \begin{pmatrix} \lambda_i & 1 & 0 & \dots & 0 \\ 0 & \lambda_i & 1 & \dots & 0 \\ 0 & 0 & \lambda_i & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & \lambda_i \end{pmatrix}$$

Each summand is called a **Jordan block**.

Proof. As an F[T]-module, decompose $V \cong \bigoplus_{i=1}^n F[T]/(T-\lambda_i)_i^j$, and choose as a basis for each summand $1, T-\lambda_i, \ldots, (T-\lambda_i)_{i-1}^{j-1}$.

Corollary 6.16 (Diagonalization Theorem). A matrix is diagonalizable iff its Jordan Canonical Form is diagonal iff its minimal polynomial is separable.

Proof. By uniqueness of the Jordan Canonical Form the first statement is true, and since the minimal polynomial is the LCM of the minimal polynomials of the Jordan blocks, so it must have distinct roots. \Box

Proposition 6.17. Commuting diagonalizable endomorphisms A, B are simultaneously diagonalizable.

Proof. If v has eigenvalue λ for A, then $BAv = ABv = \lambda Av$, so B's λ -eigenspace is A-invariant, so we can simultaneously diagonalize.

18

7. BILINEAR MAPS

Lemma 7.1. If $f: \operatorname{Sym}^2 V \to F$ is a map of F-vector spaces which is nontrivial, and $\operatorname{char}(F) \neq 2$, there is an element so that $f(v \otimes v) \neq 0$.

Proof. There is some v_1, v_2 such that $f(v_1 \otimes v_2) \neq 0$. Now WLOG, $v_1 \neq v_2$ $f(v_1 \otimes v_1) = f(v_2 \otimes v_2) = 0$, so we have $\sum_{1 \leq i,j \leq 2} f(v_i \otimes v_j) = f((v_1 + v_2) \otimes (v_1 + v_2)) \neq 0$

Proposition 7.2 (Decomposition Theorem). If $f: \operatorname{Sym}^2 V \to F$ is a map of F-vector spaces which is nontrivial, for any element $v \in V$ so that $f(v \otimes v) \neq 0$, $V \cong V' \oplus vF$, where V' consists of vectors f-orthogonal to v.

Proof.
$$\pi(x) = x - v \frac{f(v \otimes x)}{f(v \otimes v)}$$
 is a projection onto V' .

Corollary 7.3 (Graham-Schmidt Theorem). If $f: \operatorname{Sym}^2 V \to F$ is a map of F-vector spaces, V is finite dimensional, and $\operatorname{char}(F) \neq 2$, then V has an f-orthogonal basis. Hence f is represented by a diagonal matrix.

Proof. If f is 0, any basis will do. If not, by induction, Lemma 7.1 and Proposition 7.2 we are done.

Corollary 7.4 (Sylvester's Law of Inertia). If $f : \operatorname{Sym}^2 V \to \mathbb{R}$ is a map of \mathbb{R} -vector spaces, V is finite dimensional, then f is represented by a unique matrix of the form $I_n \oplus -I_m \oplus 0_r$. (Congruent real symmetric matrices have the same rank and signature).

Proof. Once we have diagonalized a matrix representing f from Corollary 7.3, we can scale each diagonal by a square, giving 1, -1, 0. This form is unique as V decomposes into V_+, V_-, V_0 where V_+ is the span of vectors v with $f(v \otimes v) = 1$ and similarly for the other two. By orthogonality, f is positive definite in V_+ and negative definite in V_- , so these subspaces, and hence the rank and signature, are invariant.

Proposition 7.5 (Sylvester's Criterion). A symmetric real matrix A is positive definite iff the principle minors are positive, and negative iff the odd principle minors are negative and the even ones positive.

Proof. If A is positive definite, it is of the form P^tP for invertible P, so $\det(A) = \det(P)^2 > 0$. Every principle submatrix is positive definite on the corresponding subspace so the principle minors are positive. Conversely if A has positive principle minors, we induct. Write $A = BA'B^t$,

$$A = \begin{pmatrix} A_0 & a \\ a^t & \alpha \end{pmatrix}, B = \begin{pmatrix} I_{n-1} & 0 \\ a^t A_0^{-1} & 1 \end{pmatrix}, A' = \begin{pmatrix} A_0 & 0 \\ 0 & \alpha - a^t A_0 a \end{pmatrix}$$

So $\alpha - a^t A_0 a$ is positive by the fact that the determinant is positive and A_0 is positive definite by induction, so we are done. For the negative definite case note A is negative definite iff -A is positive definite.

Corollary 7.6 (Hessian Test). A critical point c of a C^2 function $f : \mathbb{R}^n \to \mathbb{R}$ is a local maximum iff the Hessian matrix $(\partial_i \partial_j f(c))$ is negative definite, and a local minimum iff the Hessian matrix is positive definite.

Proof. If this matrix is positive definite, then by the Taylor expansion, f(x-c)-f(c) is closely approximated by positive definite quadratic function, so is larger than 0 in a deleted neighborhood of 0. The corresponding argument yields the other result.

8. Inner Product Spaces

Proposition 8.1. Every endomorphism $T: V \to V$ on a finite dimensional inner product space V has a unique adjoint T^* so that $(Tu, v) = (u, T^*v)$.

Proof. Choose an orthonormal basis, yielding an isometry with the standard inner product space on \mathbb{C}^n . Now if T is represented by the matrix A, T^* is represented by \bar{A}^t as we have $(Au, v) = (Au)^t \cdot \bar{v} = u^t \bar{A}^t v = (u, \bar{A}^t v)$.

Proposition 8.2. If V is a finite dimensional inner product space, T a linear operator, $\ker(T) = T^*V^{\perp}$, $(W^{\perp})^{\perp} = W, W \subset V$.

Proof. For the first part, $v \in \ker(T)$ iff $0 = (Tv, u) = (v, T^*u)$ for all u, and the second part follows from the fact that we can have an orthonormal basis for W, and extend it to an orthonormal basis for V, the remaining vectors making up a basis for W^{\perp} .

Definition 8.3. An endomorphism is **normal** if it commutes with its adjoint, and is **self-adjoint** if it is its own adjoint.

Theorem 8.4. A normal endomorphism T on a finite dimensional vector space V satisfies $||Tv|| = ||T^*v||$, $\ker T = \ker T^*$, $TV = T^*V$, $T - \lambda 1_V$ is normal, $Tv = \lambda v \leftrightarrow T^*v = \bar{\lambda}v$, and if u, v are eigenvectors of T with different eigenvalues, then (u, v) = 0

Proof. For the first, we have $(Tv, Tv) = (v, T^*Tv) = (v, (T^*)^*T^*v) = (T^*v, T^*v)$. The second immediately follows, and the third follows from the second and Proposition 8.2. For the fourth we have $(T - \lambda 1_V)(T - \lambda 1_V)^* = T^*T - \bar{\lambda}T - \lambda T^* + |\lambda|^2 1_V = (T - \lambda 1_V)^*(T - \lambda 1_V)$. The fifth follows from the fourth and the second, and the last follows from the fact that $\lambda_1(u, v) = (Tu, v) = (u, T^*v) = \lambda_2(u, v)$.

Proposition 8.5. If T is an endomorphism on a finite dimensional vector space V and $||Tv|| = ||T^*v||$ then T is normal.

Proof.
$$(T(u+v), T(u+v)) = (T^*(u+v), T^*(u+v)) \implies (Tu, Tv) = (T^*u, T^*v) \implies (u, TT^*v) = (u, T^*Tv).$$

Theorem 8.6 (Spectral Theorem for \mathbb{C}). An endomorphism T of an finite dimensional complex inner product space V is normal iff there is an orthonormal basis of eigenvectors, ie. there is a diagonal matrix representing it.

Proof. If there is an orthonormal basis of eigenvectors $v_1, \ldots v_n$, T with respect to this basis is diagonal, so commutes with its adjoint. Conversely if T is normal, its minimal polynomial has a root so it has a nontrivial eigenvector v. Then by Theorem 8.4, if u is in the orthogonal complement, $(Tu, v) = (u, T^*v) = \bar{\lambda}(u, v) = 0$, so T acts within the orthogonal complement and we can repeat until we get an orthonormal basis of eigenvectors.

Lemma 8.7. The eigenvalues of a self-adjoint endomorphism T on a vector space V are real.

Proof. This follows from the fifth part of Theorem 8.4.

Theorem 8.8 (Spectral Theorem for \mathbb{R}). An endomorphism T of a finite dimensional real inner product space V is self-adjoint iff there is an orthonormal basis of eigenvectors, ie. there is a diagonal matrix representing it.

Proof. If there is an orthonormal basis of eigenvectors, T is represented by a diagonal matrix, which is self-adjoint. Conversely if T is self-adjoint, we can tensor with $\mathbb C$ and find an eigenvector, but its eigenvalue is real, and T is real, so its conjugate is an eigenvector as well, but then either their sum is a real eigenvector or it is 0 in which case we can multiply our original eigenvector by i to get a real eigenvector. In either case, there must be a real eigenvector, so by Theorem 8.4 T acts within the orthogonal complement again, so we can repeat until we get an orthonormal basis of eigenvectors.

9. Geometry of Mappings

Theorem 9.1 (Inverse Mapping Theorem). If $f : \mathbb{F}^n \to \mathbb{F}^n$ is \mathcal{C}^1 near a point a and f'(a) is invertible near the origin, then f is a \mathcal{C}^1 diffeomorphism near a.

Proof. WLOG, a = f(a) = 0, f'(a) = I via an affine transformation. Now as r(x) = f(x) - x is \mathcal{C}^1 near the origin and r'(0) = 0, we have $||r'(x)||_o \le 1/2$ near the origin. Then by Corollary 4.7 we have $||r(b) - r(a)||_2 \le \frac{1}{2}||b - a||_2$ near the origin which gives

$$||f(b) - f(a)|| = ||f(b) - f(a)|| + \frac{1}{2}||b - a|| - \frac{1}{2}||b - a||$$

$$\ge ||f(b) - f(a)|| + ||r(b) - r(a)|| - \frac{1}{2}||b - a|| \ge \frac{1}{2}||b - a||$$

Hence the map is injective, and by Theorem 3.13, this is a local homeomorphism near the origin. Now let g be the local inverse of f near 0, we would like to show g is differentiable at 0. If f(x) = y and f(x + h) = y + k, then as f is differentiable,

$$f(x+h) - f(x) = f'(x)h + \epsilon_f(h) \implies k = f'(x)(g(y+k) - g(y)) + \epsilon_f(h)$$
$$\implies g(y+k) - g(y) = f'(x)^{-1}k - f'(x)^{-1}\epsilon_f(h)$$

and as $||f'(x)^{-1}||_o$ is bounded near 0, it suffices to show $\frac{||\epsilon_f(h)||}{||k||} \to 0$ as $k \to 0$. Indeed, we have

$$\frac{\|\epsilon_f(h)\|}{\|k\|} = \frac{\|\epsilon_f(h)\|}{\|h\|} \frac{\|h\|}{\|k\|} = \frac{\|\epsilon_f(h)\|}{\|h\|} \frac{\|h\|}{\|f(x+h) - f(x)\|} \le 2 \frac{\|\epsilon_f(h)\|}{\|h\|} \frac{\|h\|}{\|h\|}$$

which goes to 0 as $k \to 0$. Thus g is differentiable and since its derivative is the inverse of f', g is C^1 .

Theorem 9.2 (Decomposition Theorem). If a map $f : \mathbb{R}^n \to \mathbb{R}^n$ is \mathcal{C}^1 near a point a and $\det(f'(a)) \neq 0$, then near a, f is the composite of \mathcal{C}^1 diffeomorphisms ϕ_i that only change one variable.

Proof. WLOG, a = f(a) = 0. We say that f is of type r if f doesn't change at least r-1 coordinates. By induction it suffices to show that if f is type r, then there is a C^1 diffeomorphism ϕ so that $f \circ \phi$ is type r+1, so we assume f fixes the first r-1 coordinates, denoting these x_I , and denoting the last n-r coordinates x_{II} . Then after some relabeling f' looks like

$$\begin{pmatrix} I_{r-1} & 0 & 0 \\ \partial_I f_r & \partial_r f_r & \partial_{II} f_r \\ \partial_I f_{II} & \partial_r f_{II} & \partial_{II} f_{II} \end{pmatrix}$$

Since f'(x) is invertible near 0, we can assume $\partial_r f_r \neq 0$ near by relabeling the f_i . Then we can define ψ near 0 as the function that is f_r on the x_r coordinate and the identity on all other coordinates. Its derivative looks like

$$\begin{pmatrix} I_{r-1} & 0 & 0 \\ \partial_I f_r & \partial_r f_r & \partial_{II} f_r \\ 0 & 0 & I_{n-r} \end{pmatrix}$$

so is invertible and by Theorem 9.1 has a local inverse ϕ , which is what we want. \square

Note that the m = n case of the next theorem is the Inverse Mapping Theorem.

Theorem 9.3. If $f: \mathbb{R}^m \to \mathbb{R}^n$, $m \geq n$ is C^1 near a point a with f'(a) rank n, then there is a C^1 diffeomorphism $\phi: \mathbb{R}^m \to \mathbb{R}^m$ near a so that $f \circ \phi - f(a)$ is a linear map near a.

Proof. WLOG, a = 0, f(a) = 0. We label the first n coordinates of \mathbb{R}^m x_I and the last n - m x_{II} . Then f' looks like

$$(\partial_I f_I \quad \partial_{II} f_I)$$

and we define $\psi : \mathbb{R}^m \to \mathbb{R}^m$ near 0 as f_I on the first n coordinates and x_{II} on the rest. Its derivative looks like

$$\begin{pmatrix} \partial_I f_I & \partial_{II} f_I \\ 0 & I_{m-n} \end{pmatrix}$$

so by possibly reordering coordinates we can assume it is invertible and we can use Theorem 9.1 to locally make an inverse $\phi : \mathbb{R}^m \to \mathbb{R}^m$ that fixes the coordinates x_{II} . Now we have

$$\begin{pmatrix} x_I \\ x_{II} \end{pmatrix} = (\psi \circ \phi) \begin{pmatrix} x_I \\ x_{II} \end{pmatrix} = \begin{pmatrix} (f \circ \phi)(x) \\ x_{II} \end{pmatrix}$$

so indeed $f \circ \phi$ is locally linear.

Theorem 9.4 (Implicit Function Theorem). If $f: \mathbb{R}^m \to \mathbb{R}^n$, m > n is \mathcal{C}^1 near 0 and $\det(\partial_I f(0)) \neq 0$, then for a small cell $I^m = I_I \times I_{II}$ near 0, there is a \mathcal{C}^1 function $h: I_{II} \to I_I$ so that $f(x_I, x_{II}) = f(0)$ iff $x_I = h(x_{II})$.

Proof. By Theorem 9.3 we locally have a C^1 map $\phi : \mathbb{R}^m \to \mathbb{R}^n$ that is a function g on the first n coordinates and the identity on the last m-n coordinates such that $f \circ \phi$ is the linear map $(x_I, x_{II}) \mapsto (x_I)$. Then we define $h(x_{II}) = g(0, x_{II})$. Now locally f(x) = 0 iff $x = \phi(y)$ and $(f \circ \phi)(y) = 0$ iff $x = \phi(0, y_{II})$ iff $x_I = h(y_{II}) = h(x_{II})$. \square

Corollary 9.5. Locally a C^1 k-submanifold of \mathbb{R}^n is given by implicit C^1 functions in k variables.

Proof. If V is such a submanifold, after a local change of coordinates at a point a, it is a linear subspace of \mathbb{R}^n . We can collapse this subspace, which is \mathcal{C}^1 and by Theorem 9.4 the kernel composite of this with the local change of coordinates is given by implicit functions in k variables. Conversely if V is given locally by implicit functions in k variables, those implicit functions are a \mathcal{C}^1 diffeomorphism to a linear embedding of \mathbb{R}^k in \mathbb{R}^n .

Note that the n = m = r case of the Rank Theorem below is the Inverse Mapping Theorem.

Theorem 9.6 (Rank Theorem). If $f: \mathbb{R}^m \to \mathbb{R}^n$ is C^1 near a and f'(x) is rank r near a, then there are C^1 maps $\phi: \mathbb{R}^m \to \mathbb{R}^m$ defined near a and $\theta: \mathbb{R}^n \to \mathbb{R}^n$ defined near f(a) such that $\theta \circ f \circ \phi$ is a linear map of rank r.

Proof. WLOG, a = 0, f(a) = 0. By relabeling coordinates in the domain and range, we may assume that the principle $r \times r$ submatrix of f'(0) is invertible. Let us label the first r coordinates of \mathbb{R}^m x_I and the last m - r x_{II} . Then we can consider the map $\psi : \mathbb{R}^m \to \mathbb{R}^m$ that is f on x_I , and the identity on x_{II} . By hypothesis, $\psi'(0)$ is invertible, so by the Inverse Mapping Theorem we can let ϕ be its local inverse, which is g on x_I and the identity on x_{II} . Now we can call the first r coordinates of \mathbb{R}^n y_I and the last n - r y_{II} . Now $h = f \circ \phi$ is the map that is x_I on the first r coordinates, and h_{II} on the last n - r. h' looks like

$$\begin{pmatrix} I_r & 0 \\ \partial_I h_{II} & \partial_{II} h_{II} \end{pmatrix}$$

but since it is rank r near a by the chain rule, we must have $\partial_{II}h_{II} = 0$, so h only depends on x_I . Now we can define $\theta : \mathbb{R}^n \to \mathbb{R}^n$ near 0 as the identity on the first r coordinates, and $y_{II} - h_{II}(y_I)$ on the last n - r. Then $\theta'(0)$ is rank n, and $\theta \circ h$ is x_I on the first r coordinates and 0 on the last n - r, which is linear.

Corollary 9.7. If $f: \mathbb{R}^m \to \mathbb{R}^n$ is C^1 near a point a, and f'(x) is rank r near a, then the image of a small neighborhood around a is a C^1 r-submanifold of \mathbb{R}^n , and the preimage of f(a) is a C^1 m - r-submanifold if f is restricted close enough to a.

Proof. Use Theorem 9.6 to obtain ϕ and θ . Now after applying these \mathcal{C}^1 diffeomorphisms, the preimage of f(a) is the kernel of a linear map, so is an open subset of \mathbb{R}^{m-r} . Similarly, the image of a neighborhood of a is an open subset of \mathbb{R}^r .

Corollary 9.8 (Lagrange Multipliers). If $g : \mathbb{R}^n \to \mathbb{R}$ is C^1 in an open set U, where C^1 r-submanifold V that is the locus of $f_i, 1 \leq n - r$ in U, any extremum c of g in V must satisfy $\nabla g(c) = \sum_{1}^{n-r} \lambda_i \nabla f_i(c)$.

Proof. Note that for any parameterized curve ϕ on V that sends 0 to c, we have $f_i(\phi(c)) = 0$, so by the chain rule, $\nabla f_i(c) \cdot \phi'(0) = 0$, so the tangent space is exactly the space perpendicular to the ∇f_i . Now in order to have an extremum of g at c on V, we need g'(c) = 0, but then for any parameterized curve ϕ on V sending 0 to x, again we have $g(\phi(c)) = 0$, so by the chain rule, $\nabla g(c)$ lies in the tangent space, so is a linear combination of the $\nabla f_i(c)$.

10. Integration

Lemma 10.1. A countable union $\cup_i U_i$ of measure 0 sets is measure 0.

Proof. For any $\epsilon > 0$, cover each U_i with countably many cells summing to size $\leq \frac{\epsilon}{2^{i+1}}$.

Theorem 10.2 (Riemann-Lebesgue Theorem). A bounded function f on a closed cell Δ is Riemann integrable $(\int_{*\Delta} f = \int^{*\Delta} f)$ iff f is continuous almost everywhere.

Proof. Let M be the bound for f. The set of points with oscillation $\leq \epsilon$ is open, so the set of discontinuities is compact, and since f is discontinuous on a set of measure 0, by compactness this is actually content 0.

If f is continuous almost everywhere, for any $\epsilon > 0$, choose a partition \mathcal{P} such that f varies by at most $\frac{\epsilon}{2|\Delta|}$ in each cell where f is continuous, and so that the the discontinuous points are covered by cells with total content less than $\frac{\epsilon}{2M|\Delta|}$. Then we have $S^*(f,\mathcal{P}) = S(f,\mathcal{P}) = \sum_{\alpha \in \Delta} \alpha_{\alpha}(\Delta)|\Delta| \leq \frac{\epsilon}{2} + \frac{\epsilon}{2} = \epsilon \cos \int_{-\infty}^{\infty} f = \int_{-\infty}^{+\Delta} f$

we have $S^*(f, \mathcal{P}) - S_*(f, \mathcal{P}) = \sum_{\Delta \in \mathcal{P}} o_f(\Delta) |\Delta| < \frac{\epsilon}{2} + \frac{\epsilon}{2} = \epsilon$ so $\int_{*\Delta} f = \int_{*\Delta}^{*\Delta} f$. Conversely if f is not continuous almost everywhere, by Lemma 10.1 there is an ϵ such that the set of points with oscillation $\geq \epsilon$ cannot be covered by cells of size smaller than δ , so $\sum_{\Delta \in \mathcal{P}} o_f(\Delta) |\Delta|$ is larger than than $\epsilon \delta$.

Note for some of the following theorems the conditions on the function may be made weaker, ie. it can be just continuous, differentiable on the interior, with a bounded and almost everywhere continuous derivative.

Theorem 10.3 (Second Fundamental Theorem of Calculus). If $f : \mathbb{R} \to \mathbb{R}$ is C^1 near [a, b], then $\int_{[a, b]} f' = f(b) - f(a)$.

Proof. For any partition \mathcal{P} we have by Proposition 4.2 that for each $[x,y] = \Delta \in \mathcal{P}$, f'(c)(y-x) = f(y) - f(x) for a $c \in \Delta$, and as this is true for any partition, $\int_x f' \leq f(b) - f(a) \leq \int_x^* f'$, so by Theorem 10.2 we are done.

Corollary 10.4 (Integration by Parts). If f, g are C^1 near an interval [a, b], then $\int_{[a,b]} fg' = f(b)g(b) - f(a)g(a) - \int_{[a,b]} f'g$

Proof. This follows from linearity, Corollary 4.4, and Theorem 10.3. \Box

Theorem 10.5 (First Fundamental Theorem of Calculus). If $f : \mathbb{R} \to \mathbb{R}$ is continuous on the interval [a,b], then $F(x) = \int_a^x f$ is continuous on the interval and F'(x) = f(x) in the interior.

Proof.

$$F(x+h) - F(x) - f(x)h = \int_{x}^{x+h} f(t) - f(x)dt$$

and as $|f(t) - f(x)| \le \epsilon$ for small enough h and any $\epsilon > 0$, we have

$$\int_{x}^{x+h} f(t) - f(x)dt \le \int_{x}^{x+h} \epsilon \le \epsilon h$$

Theorem 10.6 (Linearity of the Integral). If f, g are Riemann integrable and D is a Jordan domain, $\int_D (c_1 f + c_2 g) = c_1 \int_D f + c_2 \int_D g$

Proof. Scaling is obvious, so it suffices to prove $\int_D (f+g) = \int_D f + \int_D g$. To see this, choose partitions \mathcal{P}_1 and \mathcal{P}_2 so that $S_*(f,\mathcal{P}_1), S_*(g,\mathcal{P}_2)$ differ by at most $\frac{\epsilon}{2}$ from $\int_D f, \int_D g$. Then taking a common refinement \mathcal{P} , we have that $S_*(f+g)$ differs from $\int_D (f+g)$ by at most ϵ . A similar argument can be made for the upper integral. \square

Theorem 10.7 (Positivity of the Integral). If $f \ge 0$ on D, then $\int_D f \ge 0$. If $f \ge g$ on D, then $\int_D f \ge \int_D g$. Also for any f, $|\int_D f| \le \int_D |f|$.

Proof. The first is obvious by looking at any partition. The second follows from the first and Theorem 10.6. The last follows from the second by noting $|f| \ge f, -f$. \square

Theorem 10.8. For a vector valued function f, $\|\int_D f\|_2 \le \int_D \|f\|_2$.

Proof. Let u be a unit vector in the direction of $\int_D f$. Then by the Cauchy-Schwarz inequality and linearity, we have

$$\left\| \int_D f \right\|_2 = u \cdot \int_D f = \int_D u \cdot f \le \int_D \|f\|_2$$

Theorem 10.9 (Invariance of the Integral). If f, g are integrable on a Jordan domain D and differ on a set of content 0, then $\int_D f = \int_D g$.

Proof. It suffices to show a function nonzero on a set of content 0 has integral 0, but this is true by definition of content 0, and the fact that the function must be bounded. \Box

Theorem 10.10 (Additivity of the Integral). If f is integrable on D, E, Jordan domains whose intersection is content 0, then $\int_{D \cup E} f = \int_D f + \int_E f$.

Proof. If χ denotes the characteristic function, then $\int_{D \cup E} f = \int_{D \cup E} (\chi_D f + \chi_E f) = \int_D f + \int_E f$, where we have ignored the boundary as it is content 0.

Theorem 10.11 (Fubini's Theorem). If f is integrable in the product cell $\Delta = \Delta_I \times \Delta_{II}$, and the functions $\int_{*\Delta_I} f, \int_{\Delta_I}^* f$ are integrable, then $\int_{\Delta} f = \int_{\Delta_{II}} \int_{*\Delta_I} f = \int_{\Delta_{II}} \int_{*\Delta_I} f$.

Proof. Suppose we have a partition \mathcal{P} that is the product of the partitions $\mathcal{P}_I, \mathcal{P}_{II}$. Then we have

$$\sum_{\Delta \in \mathcal{P}} \inf_{x \in \Delta} f(x) |\Delta| = \sum_{\Delta_{II} \in \mathcal{P}_{II}} \sum_{\Delta_{I} \in \mathcal{P}_{I}} \inf_{x \in \Delta_{I} \times \Delta_{II}} f(x) |\Delta_{I}| |\Delta_{II}|$$

$$\leq \sum_{\Delta_{II} \in \mathcal{P}_{II}} \inf_{x_{II} \in \Delta_{II}} \sum_{\Delta_{I} \in \mathcal{P}_{I}} \inf_{x_{I} \in \Delta_{I}} f(x_{I}, x_{II}) |\Delta_{I}| |\Delta_{II}|$$

$$\leq \sum_{\Delta_{II} \in \mathcal{P}_{II}} \inf_{\substack{x_{II} \in \Delta_{II} \\ x_{II} \in \Delta_{II}}} \int_{*\Delta_{I}} f(x_{I}, x_{II}) |\Delta_{II}|$$

$$\leq \sum_{\Delta_{II} \in \mathcal{P}_{II}} \sum_{\Delta_{I} \in \mathcal{P}_{I}} \sup_{\substack{x_{II} \in \Delta_{II} \\ x_{II} \in \Delta_{I}}} \sup_{x_{I} \in \Delta_{I}} f(x_{I}, x_{II}) |\Delta_{I}| |\Delta_{II}| \leq \sum_{\Delta \in \mathcal{P}} \sup_{x \in \Delta} f(x) |\Delta|$$

showing that $\int_{*\Delta} f \leq \int_{\Delta_{II}} \int_{*\Delta_{I}} f \leq \int_{\Delta}^{*} f$ but as f is integrable these are equalities. The other equality comes from dualizing the argument.

Theorem 10.12. If f_n is a uniformly convergent sequence of integrable functions in a cell Δ , the limit f is integrable and $\lim_{n\to\infty} \int_{\Delta} f_n = \int_{\Delta} f$.

Proof. To see f is integrable, note that the union of the discontinuities of the f_n is measure 0 by Lemma 10.1, so by Theorem 3.16 f is continuous away from these points. $|\int_{\Delta} f - \sum_{1}^{n} f_{n}| \leq \int_{\Delta} |f - \sum_{1}^{n} f_{n}| \to 0$ as $n \to \infty$, so this concludes the proof.

Proposition 10.13. Any open set $U \subset \mathbb{R}^m$ is the union of countably many open Jordan domains D_i with $\bar{D}_i \subset D_{i+1}$ and the D_i composed of interiors of unions of cells of a partition.

Proof. Intersect U with a ball radius n, partition \mathbb{R}^m into cells of size $\frac{1}{2^n}$ and take the interior of the union of the pieces whose boundary is completely contained inside U as D_n . To make sure $\bar{D_n} \subset D_{n+1}$, note that the boundary of D_n is compact, so is eventually covered by another D_i , so we can take a nice enough subsequence. \square

Lemma 10.14. If k < n and V is a C^1 k-submanifold of \mathbb{R}^n , then V is measure 0.

Proof. By Lemma 10.1 it suffices to do this locally, and by Corollary 9.5 V is locally given by implicit C^1 functions g_i , and so locally the derivatives are uniformly continuous, and so for any we can cover V in the plane of the variables that the g_i are a function of finitely many by cells of height arbitrarily small.

Theorem 10.15 (Change of Variables). If D, E are open sets of \mathbb{R}^n , and $\phi: D \to E$ is a C^1 diffeomorphism, then if f has a finite improper integral on E, then $(f \circ \phi)|\det \phi'|$ has a finite improper integral on D, moreover $\int_D (f \circ \phi)|\det \phi'| = \int_E f$.

Proof. We will make a series of reductions of this problem. If E_i is a sequence of Jordan domains covering E of the sort in Proposition 10.13 with $\bar{E}_i \subset E_{i+1}$, then $\phi^{-1}(E_i)$ is a sequence of Jordan domains (the boundaries are submanifolds by Corollary 9.7 hence are measure 0 by Lemma 10.14) with $\phi^{-1}(\bar{E}_i) = \overline{\phi^{-1}(E_i)} \subset \phi^{-1}(E_{i+1})$ covering E, so it suffices to show this where f is a positive function in a cell Δ in E. By passing to refinements (Theorem 10.10) and applying Theorem 9.2, Lemma 3.14, and the chain rule, it suffices to prove this when ϕ changes only 1 variable on a sufficiently small cell Δ .

Finally to reduce to when f is a constant on the cell, suppose this has been proven, and for any partition \mathcal{P} of the cell Δ , define f^* as f on the boundary of each $\Delta' \in \mathcal{P}$ and $\sup_{\Delta'} f$ on the inside. Then we have:

$$S^*(f, \mathcal{P}) = \int_{\Delta} f^* = \int_{\phi^{-1}\Delta} (f^* \circ \phi) |\det \phi'| \ge \int_{\phi^{-1}\Delta} (f \circ \phi) |\det \phi'|$$

so $\int_{\Delta}^{*} f \geq \int_{\phi^{-1}\Delta} (f \circ \phi) |\det \phi'|$. Dually we get $\int_{*\Delta} f \leq \int_{\phi^{-1}\Delta} (f \circ \phi) |\det \phi'|$.

Now we consider the case when f is constant (even f=1 suffices) and will reduce to the 1-dimensional case. We label the coordinates in $\phi^{-1}(\Delta)$ x_1 and x_{II} and in Δ y_1 and y_{II} to distinguish the coordinate ϕ changes. ϕ' looks like

$$\begin{pmatrix} \partial_1 \phi_1 & \partial_{II} \phi_1 \\ 0 & I_{n-1} \end{pmatrix}$$

so det $\phi' = \partial_1 \phi_1$. Now from the 1-dimensional case and Theorem 10.11, we get

$$\int_{\Delta} 1 = \int_{y_{II} \in \Delta_{II}} \int_{y_1 \in \Delta_1} 1 = \int_{x_{II} \in \Delta_{II}} \int_{x_1 \in \phi^{-1} \Delta_1} |\partial_1 \phi_1|$$
$$= \int_{x_{II} \in \Delta_{II}} \int_{x_1 \in \phi^{-1} \Delta_1} |\phi'| = \int_{\phi^{-1} \Delta} |\phi'|$$

Finally for the case of 1 dimension, if $\phi:[a,b]\to [c,d]$ is our function, then $\int_{[c,d]} f = F(d) - F(c)$ by Theorem 10.5, and since $(F\circ\phi)' = (f\circ\phi)\phi'$ and WMA $\phi'>0$ as ϕ is a \mathcal{C}^1 diffeomorphism (one treats the $\phi'<0$ case similarly), we have $\int_{[a,b]} (f\circ\phi)|\phi'| = \int_{[a,b]} (F\circ\phi)' = F(d) - F(c)$.

11. Line Integrals

Lemma 11.1. If $\psi : [a, b] \to \mathbb{R}^n$ is a rectifiable curve and equivalent to $\phi : [c, d] \to \mathbb{R}^n$, then ϕ is rectifiable.

Proof. Let $h:[a,b] \to [c,d]$ be the map of equivalence. Then for any partition \mathcal{P} of $[a,b], h(\mathcal{P})$ is a partition of [c,d] yielding the same lengths.

Lemma 11.2. If $\psi : [a,b] \to \mathbb{R}^n$ is a rectifiable curve, then $L(\psi) = L(\psi|_{[a,c]}) + L(\psi|_{[c,b]})$ for any $c \in [a,b]$, and $L(\psi) \ge \|\psi(b) - \psi(a)\|_2$.

Proof. For the first one, just add in the point c to any partition of [a, b]. For the second, look at the trivial (initial) partition.

Lemma 11.3. If $\psi : [0,1] \to \mathbb{R}^n$ is a rectifiable curve, then $s : [0,1] \to [0,L(\psi)]$, $s(t) = L(\psi|_{[0,t]})$ is continuously monotonically increasing, and is constant on a subinterval [a,b] iff ψ is.

Proof. Since L is non-negative, by Lemma 11.2 s is monotonic. If ψ is constant, certainly s is by the same Lemma, and conversely if ψ is not constant, then since there is a nonzero partition, s cannot be constant.

For continuity, for any $\epsilon > 0$ and $t_0 \in [0,1]$ we have $|s(t) - s(t_0)| = L(\psi|_{[t_0,t]}) = (L(\psi|_{[t_0,t]}) - L(\psi|_{[t_0,t]}, \mathcal{P})) + L(\psi|_{[t_0,t]}, \mathcal{P})$ and now we choose a \mathcal{P} to bound this. In particular, we have our partition of [0,1] be less than $\frac{\epsilon}{2}$ away from $L(\psi)$ and we assume that f varies at most $\frac{\epsilon}{2}$ in each subinterval, which is at most δ in length. As long as $t \in (t_0 - \delta, t_0 + \delta)$, we have then $(L(\psi|_{[t_0,t]}) - L(\psi|_{[t_0,t]}, \mathcal{P})) + L(\psi|_{[t_0,t]}, \mathcal{P}) \leq \epsilon$, giving continuity.

Proposition 11.4. A C^1 parameterized curve $\psi : [0,1] \to \mathbb{R}^n$ is rectifiable, its arc length given by $L(\psi) = \int_{[0,1]} \|\psi'\|_2$.

Proof. For any partition \mathcal{P} we have

$$L(\psi, \mathcal{P}) = \sum_{[a,b]\in\mathcal{P}} \|\psi(b) - \psi(a)\|_2 = \sum_{[a,b]\in\mathcal{P}} \left\| \int_{[a,b]} \psi' \right\|_2 \le \sum_{[a,b]\in\mathcal{P}} \int_{[a,b]} \|\psi'\|_2 = \int_{[0,1]} \|\psi'\|_2$$

so ψ is rectifiable. For the second part, if h > 0 is small, u is the unit vector in the direction of $\psi'(t)$ and τ comes from Theorem 4.2 we have from Cauchy-Schwarz inequality and Lemma 11.2:

$$|u \cdot \psi'(\tau)| = \frac{1}{h} |u \cdot (\psi(t+h) - \psi(t))| \le \frac{1}{h} (s(t+h) - s(t)) \le \frac{1}{h} \int_{[t,t+h]} ||\psi'||_2$$

Letting $h \to 0$ on the left we get $\|\phi'(t)\|_2$, and on the right we also get $\|\phi'(t)\|_2$ by Theorem 10.3, so the middle, which is s'(t), must be that (we also treat h < 0 similarly).

The arc length gives a natural parameterization of a curve for integration of a function, but this is unnecessary for line integration over a vector field. A vector field should be thought of as a section of the tangent bundle.

Theorem 11.5. If $f: \mathbb{R}^n \to \mathbb{R}^n$ is a vector field continuous near $\phi([a,b])$ where ϕ is a \mathcal{C}^1 parameterized curve with $\phi' \neq 0$, then if τ is the unit tangent vector and ψ is arc length parameterization, $\int_{\psi} (f \cdot \tau) ds = \int_{[a,b]} (f \circ \phi) \phi'$.

Proof. If ψ denotes the parameterization by arc length, and $\phi' \neq 0$, by Proposition 11.4 s is a \mathcal{C}^1 diffeomorphism with ψ . Moreover, by the chain rule, $\phi' = (\psi \circ s^{-1})' = (\psi' \circ s^{-1})(s^{-1})'$ so we have by Theorem 10.15 $\int_{[a,b]} (f \circ \phi) \cdot \phi' = \int_{[0,L(\phi)]} (f \circ \psi) \cdot \psi' = \int_{\psi} (f \cdot \tau) ds$.

Definition 11.6. A conservative vector field $\mathbb{R}^n \to \mathbb{R}^n$ is one which is the gradient of a function $\mathbb{R}^n \to \mathbb{R}$, which is called its **potential**.

Theorem 11.7. If $f: \mathbb{R}^n \to \mathbb{R}^n$ is a conservative vector field in a connected open set U with C^1 potential h, and γ is a C^1 parameterized curve with unit tangent vector τ from a to b with $\gamma' \neq 0$, then $\int_{\gamma} f \cdot \tau ds = h(b) - h(a)$. That f is C^1 satisfies this property is also sufficient for it to be conservative.

Proof. By Theorem 11.5 we may assume γ is arc length and by the chain rule,

$$\frac{d}{ds}h = (h' \circ \gamma)\gamma' = (\nabla h \circ \gamma) \cdot \gamma' = (f \circ \gamma) \cdot \gamma' = (f \circ \gamma) \cdot \tau$$

so this follows from Theorem 10.3. For the converse, define $h(a) = \int_{\gamma_a} f \cdot \tau ds$ where γ_a is any path to a from a fixed point b. Now we can find the partial derivatives of h at a point a by integrating along a small path $a \to a + \delta$ in one component and taking the derivative. If u is the unit vector in the x_i direction which δ is in, this yields $h(a+\delta) - h(a) = \int_{\gamma_{a,a+\delta}} f \cdot \tau ds = \int_{[0,\delta]} (f(a+\delta) - f(a)) \cdot u$ which by Theorem 10.3 shows $f_i = \partial_i(h)$.

12. Exterior Algebras and Differential Forms

Definition 12.1. If M is a finite rank free R-module with an ordered basis $m_1 \dots m_n$, an element $\omega \in \bigwedge^r(M)$ is in the **reduced form** if it is written as

$$\sum_{1 \le i_1 < \dots < i_r \le n} a_{i_1 \dots i_r} m_{i_1} \wedge \dots \wedge m_{i_r}$$

We will use **multi-index notation** writing $\sum_{I} a_{I} m_{I}$ for unreduced form, and $\sum_{I}' a_{I} m_{I}$ for reduced form.

Definition 12.2. If M is a finite rank free R-module with an ordered basis $x_1 ldots x_n$, the **Hodge duality map** denoted * is defined as the linear map sending $x_I \in \bigwedge^r M$ to $\epsilon(IJ)x_J \in \bigwedge^{n-r} M$ where ϵ is the sign of the permutation, and I, J are reduced.

Lemma 12.3. The Hodge dual satisfies $*(*\omega) = (-1)^{n(n-r)}\omega$.

Proof. It suffices to show this on dx_I , that $\epsilon(IJ)\epsilon(JI) = (-1)^{n(n-r)}$, but this is obvious by moving entries in J one at a time across as transpositions.

Lemma 12.4. If $\omega, \sigma \in \bigwedge^r, \omega = \sum_I' f_I dx_I, \sigma = \sum_I' g_I dx_I$ then $\omega \wedge *\sigma = \sum_I f_I g_I dx_1 \wedge \cdots \wedge dx_n$.

Proof. All the terms in this wedge are 0 except the ones where the f_I and g_I correspond, in which case the sign is $\epsilon(IJ)\epsilon(IJ) = 1$

We will work in the smooth category of \mathcal{C}^{∞} functions for simplicity, but if one is careful it is possible to treat \mathcal{C}^r functions. We think of the **de Rham complex** $\bigwedge(D)$ of $D \subset \mathbb{R}^n$ as the exterior algebra of the $\mathcal{C}^{\infty}(D)$ -module of smooth sections of

the cotangent bundle $T^*(D)$, where $\bigwedge^r(D)$ is a free module of rank $\binom{n}{r}$. Elements of $\bigwedge^r(D)$ are called **differential r-forms**. Since here D is in \mathbb{R}^n , by choosing an orientation $x_1 \dots x_n$ of \mathbb{R}^n , we determine a basis for $\bigwedge^1(D)$, $dx_1 \dots dx_n$, namely dx_1 is the section taking every point to the projection onto the x_i coordinate (recall that we have a canonical identification of the tangent bundle with \mathbb{R}^n).

The de Rham complex is a cochain complex because it has a differential d, the exterior derivative.

Definition 12.5. The exterior derivative is defined by

$$d\left(\sum_{I} f_{I} dx_{I}\right) = \sum_{I} df_{I} dx_{I}$$

where df is the dual of f' written in the basis of the dxs, ie $df = \sum_i \partial_i dx_i$

Note that $d(x_i) = dx_i$ where x_i is the i^{th} coordinate function (indeed this is the meaning of dx_i). The cohomology of the de Rham complex is called **de Rham** cohomology. The exterior derivative is characterized by the following properties:

Lemma 12.6. The exterior derivative d is an \mathbb{R} -linear map satisfying d(f) = df, $(\omega \wedge \sigma) = d\omega \wedge \sigma + (-1)^r \omega \wedge d\sigma$ and $dd\omega = 0$ where $\omega \in \bigwedge^r(D)$, $\sigma \in \bigwedge^k(D)$. Moreover it is characterized by these properties.

Proof. d is clearly linear and satisfies the first property by definition. The second follows from the product rule with the $(-1)^r$ coming from moving the derivative of each g in the σ to the right place. For the third, it suffices to show that ddf = 0 for a function, but this is true by Theorem 4.10 since $ddf = \sum_{i,j} \partial_i \partial_j dx_i dx_j = 0$. To see that d is characterized by these properties, note that by induction $d(dx_I) = 0$, so this allows us to compute $d(f \wedge dx_I)$ as defined.

Now when we have a C^{∞} map $\phi: D \to E$, we would like to be able to pullback differential forms to get a map of cochain complexes, $\phi^*: \bigwedge(E) \to \bigwedge^r(D)$. The pullback will serve as the chain map that turns de Rham cohomology into a functor. To define this, note that we have the pushforward, a map $d\phi: TD \to TE$, and by taking its dual (transpose) and exterior algebra, this gives a map $\bigwedge(d\phi)^*: \bigwedge T^*E \to \bigwedge T^*D$. We can define the **pullback** as the composite in the diagram below:

$$\bigwedge(T^*D) \stackrel{\bigwedge(d\phi)^*}{\longleftarrow} \bigwedge(T^*E)$$

$$\downarrow^{\phi^*(\omega)} \qquad \qquad \uparrow^{\omega}$$

$$D \stackrel{\phi}{\longrightarrow} E$$

How do we compute the pullback of ω ? Well first write $\omega = \sum_I f_I dx_I$. The f_I get sent by ϕ to $f \circ \phi$, and since $(d\phi)^*$ is $\mathcal{C}^{\infty}(E)$ -linear on sections, we only need to find

out what happens to the dx_i . To do this note that $d\phi$ sends $(a, v) \mapsto (\phi(a), \phi'(a)v)$, which $dx_i(\phi(a))$ sends to $\pi_{x_i}(\phi'(a))v = \phi'_i(a)v$ which is exactly what $\sum_j \partial_j \phi_i dy_j$ does, so $dx_i \mapsto \sum_j \partial_j \phi_i dy_j$, and we can compute the pullback as

$$\phi^*(fdy_1 \wedge \cdots \wedge dy_r) = \sum_{j} (f \circ \phi) \partial_{j_1} \phi_{i_1} \dots \partial_{j_r} \phi_{i_r} dx_{j_1} \wedge \cdots \wedge dx_{j_r}$$

which is abbreviated as

$$\phi^*(f_I(y)dy_I) = f_I(\phi(y)) \sum_J \partial_J \phi_I dx_J$$

Lemma 12.7. Pullback is a contravariant functor and is a map of cochain complexes.

Proof. Let's consider the pullback of a map $\phi: D \to E$, $D \subset \mathbb{R}^m$, $E \subset \mathbb{R}^n$. The pullback is the composite of maps along the functor $\bigwedge(d(-))^*$, which is a functor by the chain rule, so it is a functor. By definition pullback is a map of algebras, and so we must check it commutes with the exterior derivative. For any function $f: E \to \mathbb{R}$ we have by the chain rule $\phi^*(df) = \phi^*(\sum_i \partial_i f dx_i) = \sum_i (\partial_i f \circ \phi)(\sum_j \partial_j \phi_i dy_j) = \sum_j \sum_i (\partial_j f \circ \phi) \partial_j \phi_i dy_j = \sum_i \partial_i (f \circ \phi) dy_i = d(f \circ \phi) = d(\phi^*(f))$. Now by linearity we only need commutativity on $\omega = f dx_I$ so

$$\phi^*(d\omega) = \phi^*(df \wedge dx_I) = \phi^*(df) \wedge \phi^*(dx_I) = d\phi^*(f_I) \wedge \phi^*(dx_I)$$
$$= d(\phi^*(f_I)\phi^*(dx_I)) = d(\phi^*(\omega))$$

Theorem 12.8 (Poincaré Lemma). The de Rham cohomology of a cell is \mathbb{R} in dimension 0 and 0 elsewhere.

Proof. We assume the cell is a unit cube. That the cohomology is \mathbb{R} in dimension 0 follows from the fact that the derivative is 0 iff the function is constant, and since the cell is connected, there the function must globally be constant. For elsewhere, we consider a closed differential r-form and show it is exact. To do this we induct on the largest k so that dx_k is used in the form. For the base case, if $\omega = f dx_1$ then $dg = \omega$ where $g = \int_0^{x_1} f$. Now to induct on k, it suffices by linearity and induction to consider a r-form of the sort $\omega = f dx_1 \wedge dx_k$ where dx_I only use $dx_1 \dots dx_{k-1}$. As it is closed, we have

$$0 = d\omega = \sum_{j=1}^{n} \partial_j f dx_j \wedge dx_I \wedge x_k$$

so $\partial_j f$ must be 0 for j > k, so f is constant in those directions, so we define $h = \int_0^{x_k} f$ so that $\partial_k h = f$, and setting $\omega' = h dx_I$ and changing ω by $(-1)^{r-1} d\sigma$ we are done by induction.

13. Integrals of Differential Forms

If we have an orientation $y_1 \dots y_n$ on \mathbb{R}^n , we define $\int_E f dy_1 \wedge \dots \wedge dy_n = \int_E f$ for $E \subset \mathbb{R}^n$.

Theorem 13.1. If $\phi: D \to E$ is a C^1 diffeomorphism between two subsets of \mathbb{R}^n , and ω is a continuous differential n-form, then $\int_E \omega = \epsilon \int_D \phi^*(\omega)$ where ϵ is the sign of the determinant of ϕ' .

Proof. If $x_1
ldots x_n$ is the orientation for D, $y_1
ldots y_n$ is the orientation for E, and $\omega = f dy_I$ we have $\int_D \phi^*(\omega) = \int_D (f \circ \phi) \det(\phi') dx_I = \int_D (f \circ \phi) \det(\phi') = \epsilon \int_E f = \epsilon \int_E \omega$ by the change of variables formula.

A singular n cell in \mathbb{R}^m is a map from an oriented cell Δ_n to \mathbb{R}^m . We will mostly consider \mathcal{C}^{∞} singular n cells. Two \mathcal{C}^{∞} singular n cells are equivalent if there is an orientation preserving \mathcal{C}^{∞} diffeomorphism between their domains that commutes with them. If we have a differential form ω in an open set containing a singular cell $\phi: \Delta_n \to \mathbb{R}^m$, we define $\int_{\phi} \omega = \int_{\Delta_n} \phi^*(\omega)$. By Theorem 13.1 this only depends on the equivalence class. Indeed given a notion of integration, the pullback may be defined as the unique form satisfying $\int_D \phi^*(\omega) = \int_{\phi(D)} \omega$ (it is dual to the pushforward of a chain).

Indeed if ϕ is a singular 1 cell with $\phi' \neq 0$, and $\omega = \sum_{i=1}^{n} f_i dx_i$, is a 1-form, then if f is the vector field with components f_i , then $\int_{\phi} \omega = \int_{[0,1]} (f \circ \phi) \cdot \phi' = \int_{\phi} f \cdot \tau ds$.

Now since we have an inner product structure on \mathbb{R}^n , we can identify covector with vectors, as a covector corresponds to a unique vector such that dotting with that vector is that covector. Thus covector fields (1-forms) may be identified with vector fields. Similarly if ϕ is a singular 2 cell with orientation t_1, t_2 , and codomain \mathbb{R}^3 with orientation x_1, x_2, x_3 , we can integrate a 2-form associated with a vector field f $\omega = f_1 dx_1 + f_2 dx_2 + f_3 dx_3$ over this cell, yielding $\int_{\phi} \omega = \int_{\Delta} f(\phi(t)) \cdot (\partial_1 \phi \times \partial_2 \phi) dt_1 \wedge dt_2$ where \times denotes cross product. If ν denotes the unit vector in the direction of $\partial_1 \phi \times \partial_2 \phi$, then ν is the **normal unit vector**, and we can write $\partial_1 \phi \times \partial_2 \phi = \nu \|\partial_1 \phi \times \partial_2 \phi\|_2$. Now we can interpret $\int_{\Delta} \|\partial_1 \phi \times \partial_2 \phi\|_2 = \int_{\phi} 1 dS$ as the surface area, and write $\int_{\phi} \omega = \int_{\phi} f \cdot \nu dS$.

We would like to integrate over chains rather than cells, so we define a **singular** n-**chain** as a finite formal sum of singular n cells. We can let $\Delta(x_1, \ldots, x_n)$ be the unit cell in \mathbb{R}^n with that orientation, and then let its boundary be defined as the singular n-1-chain

$$\partial \Delta(x_1, \dots, x_n) = \sum_{i=1}^n \sum_{\epsilon=0}^1 (-1)^{i+\epsilon} \Delta(x_1, \dots, \epsilon, \dots, x_n)$$

where ϵ denotes the part of the boundary restricting x_i to ϵ . For a singular *n*-chain ϕ , we define $\partial \phi = \phi(\partial \Delta(x_1, \ldots, x_n))$. We then have our notions of boundary and cycle for chains.

Lemma 13.2. A boundary is a cycle.

Proof. It suffices to show this on the unit cell Δ . We have

$$\partial \partial \Delta(x_1, \dots, x_n) = \sum_{i=1}^n \sum_{\epsilon=0}^1 (-1)^{i+\epsilon} \partial \Delta(x_1, \dots, \epsilon, \dots, x_n)$$

$$= \sum_{i=1}^{n} \sum_{\epsilon=0}^{1} \sum_{i=1}^{n-1} \sum_{\epsilon'=0}^{1} (-1)^{i+\epsilon+j+\epsilon'} \Delta(x_1, \dots, \epsilon', \dots, \epsilon, \dots, x_n)$$

Now for a fixed ϵ' , ϵ and unordered pair (a,b), the term $\Delta(x_1,\ldots,\epsilon',\ldots,\epsilon,\ldots,x_n)$ where ϵ' and ϵ take up the a and b slots respectively appear twice in this sum. WLOG, a < b, and so we can have both i = a, j = b - 1 or i = b, j = a. Then as i + j is a different parity for these, the terms cancel out, so this sum is 0.

Theorem 13.3 (Stoke's Theorem). If C is a C^{∞} singular n-chain and ω is a n-1-form, then $\int_{C} d\omega = \int_{\partial C} \omega$.

Proof. Since pullback commutes with exterior derivative and integration over chains is linear, it suffices to prove this on the unit cell $\Delta(x_1, \ldots, x_n)$ for a n-1-form $\omega = f dX_r$ where dX_r indicates dx_r is missing from $dX = dx_1 \wedge \cdots \wedge dx_n$. This way $d\omega = (-1)^{r-1} \partial_r f dX$. Similarly let Δ denote the cell, $\Delta_{r,e}$ denote the boundary on the r^{th} side with ϵ , Δ_r denote the cell Δ with the r dimension missing. Then from Theorem 10.11 and Theorem 10.3 we get

$$\int_{\Delta} d\omega = (-1)^{r-1} \int_{\Delta} \partial_r f dX = (-1)^{r-1} \int_{\Delta_r} \int_{\Delta(x_r)} \partial_r f$$

$$= (-1)^{r+1} \int_{\Delta_{r,1}} f(x_1, \dots, 1, \dots, x_n) + (-1)^r \int_{\Delta_{r,0}} f(x_1, \dots, 0, \dots, x_n) = \int_{\partial \Delta} f dX_r$$

Where $\int_{\partial \Delta} f dX_r$ vanishes on all other parts of the boundary as on $\Delta_{i,\epsilon}$ with $i \neq r$ the x_i part is constant, so pulling back yields the form 0.

Indeed nothing deep is going on here, one may define the exterior derivative as the map so that Theorem 13.3 holds. Note that there is a pairing between chains and forms (homology and cohomology), and thus Theorem 13.3 says simply that d is adjoint to ∂ for this pairing.

Corollary 13.4. If $H_1(U,\mathbb{Z}) = 0$ for an open set $U \subset \mathbb{R}^n$, then any closed C^{∞} differential 1-form is exact.

Proof. By Theorem 11.7 it suffices to show line integrals depend only on the start and end, but this follows from the hypothesis and Theorem 13.3. \Box

Corollary 13.5. If D is the image of a simple (orientation preserving diffeomorphism) singular C^{∞} n-chain C and $\omega = \sum_{i} f_{i} dX_{i}$ is an n-1-form, then

$$\int_{D} \sum_{i} (-1)^{i-1} \partial_{i} f_{i} = \int_{\partial \mathcal{C}} \omega$$

Proof. This follows from Theorem 13.3 and Theorem 13.1.

Corollary 13.6 (Green's Theorem). If D is the image of a simple C^{∞} singular 2-chain C, then

$$\int_{\partial \mathcal{C}} f_1 dx_1 + f_2 dx_2 = \int_D \partial_1 f_2 - \partial_2 f_1$$

Proof. This is a special case of Theorem 13.5.

Green's theorem can be used to calculate area by integrating on the boundary the form x_1dx_2 or $-x_2dx_1$ for example. For the case of $\partial \mathcal{C}$ being a curve γ , we can get $\int_{\gamma} f \cdot \nu ds = \int_{D} \partial_1 f_1 + \partial_2 f_2$ where ν is the outward pointing normal vector defined as the unit vector in the direction of $\begin{pmatrix} \partial_1 \gamma_2 \\ -\partial_1 \gamma_1 \end{pmatrix}$. Then the left side is interpreted as work done by a vector field pushing a particle, and the right is interpreted as total flow of a substance moving across the curve with velocity given by the vector field.

Corollary 13.7 (Gauss's Theorem). If D is the image of a simple C^{∞} singular 3-chain C, then

$$\int_{\partial \mathcal{C}} f_1 dx_2 \wedge dx_3 + f_2 dx_3 \wedge dx_1 + f_3 dx_1 \wedge dx_2 = \int_D \sum_{i=1}^3 \partial_i f_i$$

Proof. This is a special case of Theorem 13.5.

This can also be used to calculate volume, and can similarly be written as $\int_{\gamma} f \cdot \nu dS = \int_{D} \nabla \cdot f$, which can be interpreted as flow and amount of substance created.

Corollary 13.8 (Kelvin-Stokes Theorem). If C is a C^1 singular 2-chain, then

$$\int_{\partial \mathcal{C}} f \cdot \tau ds = \int_{\mathcal{C}} (\nabla \times f) \cdot \nu dS$$

Proof. This is a special case of Theorem 13.3.

As an interpretation of the curl, let f be a vector field in \mathbb{R}^3 , and we can integrate around a tiny circle γ_ϵ around a point a which bounds a disk D_ϵ which has its unit normal vector ν pointing in the direction of $\nabla \times f(a)$. Then by Kelvin-Stokes, $\frac{1}{\pi \epsilon^2} \int_{\gamma_\epsilon} f \cdot \tau ds = \frac{\int_{D_\epsilon} (\nabla \times f) \cdot \nu dS}{\int_{D_\epsilon} dS}.$ But as the curl is continuous, we get

$$\lim_{\epsilon \to 0} \frac{1}{\pi \epsilon^2} \int_{\gamma_{\epsilon}} f \cdot \tau ds = \|\nabla \times f(a)\|_2$$