

DEFVecSp : $1.x + y = y + x$ $2.(x + y) + z = x + (y + z)$ $3.Add.Id.0 \in V | 0 + x = x$ $4.\exists Add.Inv.(-x)|x + (-x) = 0(5).F.Dis.Lawa(x+y) = ax + ay(6).S.Dis.Law(a+b)(x) = ax + bx(7).Mul.Id.1x = x(8).(ab)x = a(bx)$ **THM1.1.13** If W is a subset of a vector space V s.t. $\mathbf{x}, \mathbf{y} \in \mathcal{W}$ and for any $a, b \in \mathbb{F}$ the vector $a\mathbf{x} + b\mathbf{y} \in W$, then W is a subspace of V . **DEFLinHull** of $S(S)$, smallest subspace of V that contains S , i.e. intersection of all subspaces of V that contain S . **THM1.2.6Span(S)** = $\langle S \rangle$. **DEF** \oplus Where W_1, W_2 are subspaces of V , then $W_1 + W_2 = W_1 \oplus W_2$ if $W_1 \cap W_2 = 0$. **DEFComplementarysubspaces** W_1 and W_2 if $V = W_1 \oplus W_2$ **THMRreplacement**: V is a vector space spanned by $S = s_1, \dots, s_m$. If $T = t_1, \dots, t_n$ is a L.I. subset of V , then $\leq m$ and $\exists S' \subset S$ having $m - n$ elements such that $T \cup S'$ spans V . **THMExtension**: W is a subspace of V If $T = t_1, \dots, t_n$ and $S = s_1, \dots, s_m$ span W and V , respectively, then $\exists S' \subset S$ having $m - n$ elements such that $T \cup S'$ is a basis for V . **DEFQuotientSpaces**: W subspace of V . The set $x + W | x \in V$ (or equivalently $[x] | x \in V$) of all cosets of W in V is denoted V/W and is called the quotient of V modulo W . **DEF** $\boxplus \boxminus$: Let W be a subspace of V . Define operations $\boxplus : V/W \times V/W \rightarrow V/W$ and $\boxminus : \mathbb{F} \times V/W \rightarrow V/W$ given by (i) $(x + W) \boxplus (y + W) = (x + y) + W$ and $a \boxminus (x + W) = (ax) + W$. These are the operations of vector addition and scalar multiplication on V/W . **CHAP2** **DEFLineartransformation** Let V and V be vector spaces over \mathbb{F} . A map $L : V \rightarrow W$ is a linear transformation from V into W if $L(ax_1 + bx_2) = aL(x_1) + bL(x_2)$ for $x_1, x_2 \in V$ and $a, b \in \mathbb{F}$ **COR2.1.17** A linear transformation is invertible if and only if it is bijective. **Prop.2.1.24** : If $V \cong W$ are isomorphic vector spaces, with isomorphism $L : V \rightarrow W$, then: (i) A linear equation holds in V iff it also holds in W : that is $\sum_{i=1}^n a_i \mathbf{x}_i = \mathbf{0}$ holds in V iff $\sum_{i=1}^n a_i L\mathbf{x}_i = \mathbf{0}$ holds in W . (ii) A set $B = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ is a basis of V iff $LB = \{L\mathbf{v}_1, \dots, L\mathbf{v}_n\}$ is a basis for W . Moreover, the dimension of V is equal to the dimension of W . (iii) The subspaces of V are in bijective correspondence with the subspaces of W . (iv) If $K : W \rightarrow U$ is any linear transformation, then the composition $KL : V \rightarrow U$ is also a linear transformation and we have $\mathcal{N}(KL) = L^{-1}\mathcal{N}(K) = \{v | L(v) \in \mathcal{N}(K)\}$ and $\mathcal{R}(KL) = \mathcal{R}(K)$ **THMF.Iso.** If V and X are vector spaces and $L : V \rightarrow X$ is a linear transformation, then $V/\mathcal{N}(L) \cong \mathcal{R}(L)$. In particular, if L is surjective, then $V/\mathcal{N}(L) \cong X$. **THM2.2.7** If V is a finite-dimensional vector space and W is a subspace of V , then $\dim(V) = \dim(W) + \dim(V/W)$ **THMRank - Nullity** Let V and W be finite-dimensional vector spaces. If $L : V \rightarrow W$ is a linear transformation then $\dim(V) = \dim \mathcal{R}(L) + \dim \mathcal{N}(L) = \text{rank}(L) + \text{nullity}(L)$. **CORSec.Iso.Thm.** Assume V_1 and V_2 are subspaces of V . Then $V_1/(V_1 \cap V_2) \cong (V_1 + V_2)/V_2$. **CORDim.Formula** If V_1 and V_2 are finite-dimensional subspaces of a vector space V , then $\dim(V_1) + \dim(V_2) = \dim(V_1 \cap V_2) + \dim(V_1 + V_2)$ **DEFSimilarMatrices** Two square matrices $A, B \in M_n(\mathbb{F})$ are said to be similar if there exists a nonsingular $P \in M_n(\mathbb{F})$ such that $B = P^{-1}AP$. **DEFBernsteinPolynomials** Given $n \in \mathbb{N}_{\geq 0}$, the Bernstein polynomials $B_j^n(x)_{j=0}^n$ of degree n are defined as $B_j^n(x) = \binom{n}{j} x^j (1-x)^{n-j}$, where $\binom{n}{j} = \frac{n!}{j!(n-j)!}$ **LEM2.5.3** For $j = 0, 1, \dots, n$ $B_j^n(x) = \sum_{i=j}^n (-1)^{i-j} \binom{n}{i} \binom{i}{j} x^i$ **THM2.5.4** For any $n \in \mathbb{N}$, the set T_n of degree n Bernstein polynomials $T_n = B_j^n(x)_{j=0}^n$ forms a basis for $\mathbb{F}[x]^n$ **DEFTrace** The trace is the sum of the elements along the main diagonal **PROP2.6.2** All of the elementary matrices are invertible. **DEFRowEquivalence** The B is said to be row equivalent to the matrix A if there exists a finite collection of elementary matrices E_1, E_2, \dots, E_n such that $B = E_1 E_2 \dots E_n A$ **DEFREF** A is REF if (i) leading coefficient of each row is strictly to the right of the previous row's leading coefficient (ii) All nonzero rows are above any zero rows and **RREF** if (iii) the leading coefficient of every row is 1 (iv) The leading coefficient of every row is the only nonzero entry in its column. **DEFPermutation** Different arrangements of a set. Even if it has an even number of inversions, odd if an odd number of inversions. Sign is 1 if even, -1 if odd. **DEFInversion** A pair $(\sigma(i), \sigma(j))$ such that $i < j$ and $\sigma(i) > \sigma(j)$. **THM2.8.7** If $A, B \in M_n(\mathbb{F})$, then $\det(AB) = \det(A)\det(B)$ **COR2.8.8** $\det(A^{-1}) = (\det(A))^{-1}$ **Cramer'sRule** If $A \in M_n(\mathbb{F})$ is nonsingular, then the unique solution to $Ax = b$ is $x = A^{-1}b = \frac{\text{adj}(A)}{\det(A)} b$. Moreover, if $A_i(b) \in M_n(\mathbb{F})$ is the matrix A with the i -th column replaced by b , then the i -th coordinate of x is $x_i = \frac{\det(A_i(b))}{\det(A)}$ **Exam2** **DEFinnerproduct** on V for $\mathbf{x}, \mathbf{y}, \mathbf{z} \in V$, $a, b \in \mathbb{F}$: (i) $\langle \mathbf{x}, \mathbf{x} \rangle \geq 0$, eq. iff $\mathbf{x} = 0$ (ii) $\langle \mathbf{x}, a\mathbf{y} + b\mathbf{z} \rangle = a\langle \mathbf{x}, \mathbf{y} \rangle + b\langle \mathbf{x}, \mathbf{z} \rangle$ (iii) $\langle \mathbf{x}, \mathbf{y} \rangle = \overline{\langle \mathbf{y}, \mathbf{x} \rangle}$ **PROP**: (i) $\langle \mathbf{x} + \mathbf{y}, \mathbf{z} \rangle = \langle \mathbf{x}, \mathbf{z} \rangle + \langle \mathbf{y}, \mathbf{z} \rangle$ (ii) $\langle a\mathbf{x}, \mathbf{y} \rangle = a\langle \mathbf{x}, \mathbf{y} \rangle$ **FunFact**: $0 \leq \langle \mathbf{x} - \lambda \mathbf{y}, \mathbf{x} - \lambda \mathbf{y} \rangle = \langle \mathbf{x}, \mathbf{x} \rangle - \langle \lambda \mathbf{y}, \mathbf{x} \rangle - \langle \mathbf{x}, \lambda \mathbf{y} \rangle + \langle \lambda \mathbf{y}, \lambda \mathbf{y} \rangle = \langle \mathbf{x}, \mathbf{x} \rangle - \overline{\lambda} \langle \mathbf{x}, \mathbf{y} \rangle - \lambda \langle \mathbf{x}, \mathbf{y} \rangle + \lambda \overline{\lambda} \langle \mathbf{y}, \mathbf{y} \rangle$ **Frobenius inner product**: $\langle A, B \rangle = \text{tr}(A^H B)$ **Orthogonal** if $\langle \mathbf{y}, \mathbf{x} \rangle = 0$. **Cauchy-Schwarz**: $|\langle \mathbf{x}, \mathbf{y} \rangle| \leq \|\mathbf{x}\| \|\mathbf{y}\|$ **PROOF**: suppose $\mathbf{u}, \mathbf{v} \neq 0$, choose $\lambda = \frac{\langle \mathbf{v}, \mathbf{u} \rangle}{\langle \mathbf{u}, \mathbf{u} \rangle}$, and $|\lambda| = 1$. Thus, $0 \leq \left\| \frac{\lambda \mathbf{u}}{\|\mathbf{u}\|} - \frac{\mathbf{v}}{\|\mathbf{v}\|} \right\|^2 = |\lambda|^2 - 2\Re\left(\frac{\lambda \mathbf{v}}{\|\mathbf{u}\| \|\mathbf{v}\|}\right) + 1 = 2 - 2 \frac{\langle \mathbf{v}, \mathbf{u} \rangle}{\|\mathbf{u}\| \|\mathbf{v}\|} \implies |\langle \mathbf{u}, \mathbf{v} \rangle| = \lambda \|\langle \mathbf{v}, \mathbf{u} \rangle\| \leq \|\mathbf{u}\| \|\mathbf{v}\|$ **PR. 3.1.23**: If \mathbf{u} is unit, $\text{proj}_{\mathbf{u}} : V \rightarrow V$ is a linear operator and: (i) $\text{proj}_{\mathbf{u}} \circ \text{proj}_{\mathbf{u}} = \text{proj}_{\mathbf{u}}$ (ii) $\mathbf{r} = \mathbf{v} - \text{proj}_{\mathbf{u}}(\mathbf{v})$ is orthogonal to $\text{span}\{\mathbf{u}\}$ (iii) $\text{proj}_{\mathbf{u}}(\mathbf{v})$ unique vector in $\text{span}\{\mathbf{u}\}$ nearest \mathbf{v} . **Angle**:

$\cos(\theta) = \frac{\langle \mathbf{x}, \mathbf{y} \rangle}{\|\mathbf{x}\| \|\mathbf{y}\|}$ Pyth: if \mathbf{x}, \mathbf{y} are orthogonal $\|\mathbf{x} + \mathbf{y}\|^2 = \|\mathbf{x}\|^2 + \|\mathbf{y}\|^2$ TH. 3.2.3:

If $(V, \langle \cdot, \cdot \rangle)$ and $\{\mathbf{x}_i\}_{i=1}^m$ is a finite orthonormal set: (i) If $\mathbf{x} = \sum_{i=1}^m a_i \mathbf{x}_i$, then $\langle \mathbf{x}_i, \mathbf{x} \rangle = a_i$ for all $i = 1, 2, \dots, m$ (ii) If $\mathbf{x} = \sum_{i=1}^m a_i \mathbf{x}_i$ and $\mathbf{y} = \sum_{i=1}^m b_i \mathbf{x}_i$, then $\langle \mathbf{x}, \mathbf{y} \rangle = \sum_{i=1}^m \bar{a}_i b_i$ (iii) If $\mathbf{x} = \sum_{i=1}^m a_i \mathbf{x}_i$, then $\|\mathbf{x}\|^2 = \sum_{i=1}^m |a_i|^2$. linear map is orthonormal if for every \mathbf{x}, \mathbf{y} $\langle \mathbf{x}, \mathbf{y} \rangle_V = \langle L\mathbf{x}, L\mathbf{y} \rangle_W$ Proj onto unit: $\text{proj}_u(\mathbf{x}) = \langle \mathbf{u}, \mathbf{x} \rangle \mathbf{u}$ THM: Q, Q_1 orthonormal: (i) $\|Q\mathbf{x}\| = \|\mathbf{x}\|$ (ii) QQ_1 is orthonormal (iii) $Q^{-1} = Q^H$ is orthonormal (iv) $Q^H A = Q Q^H = I$ (v) columns are orthonormal (vi) $|\det(Q)| = 1$ DEF: Bessel's Inequality: If $(V, \langle \cdot, \cdot \rangle)$ and $\{\mathbf{x}_i\}_{i=1}^m$ is a finite orthonormal set:

$\|\mathbf{v}\| \geq \sum_{i=1}^m |\langle \mathbf{x}_i, \mathbf{v} \rangle|^2 = \|\text{proj}_X \mathbf{v}\|^2$. DEF: Pythagorean Theorem: If $(V, \langle \cdot, \cdot \rangle)$ and $\{\mathbf{x}_i\}_{i=1}^m \subset V$ is a finite orthonormal subset with span X , for every $\mathbf{v} \in V$:

$\|\mathbf{v}\|^2 = \|\text{proj}_X(\mathbf{v})\|^2 + \|\mathbf{v} - \text{proj}_X(\mathbf{v})\|^2 = \sum_{i=1}^m |\langle \mathbf{x}_i, \mathbf{v} \rangle|^2 + \|\mathbf{v} - \sum_{i=1}^m \langle \mathbf{x}_i, \mathbf{v} \rangle \mathbf{x}_i\|^2$ Gram-Schmidt: Given $(x_1 \dots x_n)$ to get orthonormal basis $(q_1 \dots q_n)$ 1) $q_1 = \frac{x_1}{\|x_1\|}$

2) $p_1 = \text{proj}_{q_1}(x_2) = \langle q_1, x_2 \rangle q_1$, and $q_2 = \frac{x_2 - p_1}{\|x_2 - p_1\|}$ Repeat, $p_{n-1} = \langle q_1, x_n \rangle q_1 + \dots + \langle q_{n-1}, x_n \rangle q_{n-1}$, and $q_n = \frac{x_n - p_{n-1}}{\|x_n - p_{n-1}\|}$ QR: $A = QR$ where Q is orthonormal columns

of A , calculated through Gram-Schmidt. Note, $A = QR \implies Q^H A = R$. So calculate R . HyperPlane: Defined as perpendicular to a particular v , thus $\text{proj}_V(x) = x - \langle v, x \rangle v$ and the transformation is given by $I - 2 \frac{vv^H}{v^H v}$ NORMS: (i) Positivity

$\|x\| \geq 0$ with equality if and only if $x = 0$ (ii) Scale preservation $\|ax\| = |a| \|x\|$ (iii) Triangle inequality (Follows from Cauchy Schwarz) $\|x + y\| \leq \|x\| + \|y\|$ Every inner-product has norm $\|x\| = \sqrt{\langle x, x \rangle}$ Norms: $\|x\|_p = (\sum |x|^p)^{1/p}$ $\|A\|_F = \sqrt{\text{tr}(A^H A)}$

Induced Norm on Linear Transformations: $\|T\|_{V,W} = \sup_{\|x\|_V=1} \|Tx\|_W$ THM: If $T \in \mathcal{B}(X, Y)$, $S \in \mathcal{B}(Y, Z)$ then $ST \in \mathcal{B}(X, Z)$ and $\|ST\|_{X,Z} \leq \|S\|_{Y,Z} \|T\|_{X,Y}$ Remark: for $n \geq 1$ we have $\|T^n\| \leq \|T\|^n$. If $\|T\| \leq 1$ then $\|T\|^n$ approaches 0.

EX: $\|A\|_p = \sup_{x \neq 0} \frac{\|Ax\|_p}{\|x\|_p}$, $1 \leq p \leq \infty$ Young's: $ab \leq \frac{a^p}{p} + \frac{b^q}{q}$ if $\frac{1}{p} + \frac{1}{q} = 1$

Arithmetic Geometric mean: $a^\theta b^{1-\theta} \leq \theta a + (1-\theta)b$ for $0 \leq \theta \leq 1$. Holder's: if $\frac{1}{p} + \frac{1}{q} = 1$ where $1 \leq p \leq \infty$ then, $\sum |xy| \leq (\sum |x|^p)^{1/p} (\sum |y|^q)^{1/q} = \|x\|_p \|y\|_q$

(Note, $p = q = 2$ implies Cauchy Swarz) Minkowski: $\|x + y\|_p \leq \|x\|_p + \|y\|_p$ Finite Dimensional Riesz Thm: Let $L : V \rightarrow \mathbb{F}$, $\exists! y \in V$ s.t. $L(x) = \langle y, x \rangle \forall x \in V$, and $\|L\| = \|y\| = \sqrt{\langle y, y \rangle}$ Adjoint: Adjoint of L is a linear transformation s.t.

$\langle w, L v \rangle_W = \langle L^* w, v \rangle_V \forall v \in V, w \in W$. THM: Let $L : V \rightarrow W$ be finite, adjoint L^* exists and is unique. Proof: Let $L_w : V \rightarrow \mathbb{F}$ be defined by $L_w(v) = \langle w, L(v) \rangle_W$. By Riesz, $\exists! u \in V$ s.t. $L_w(v) = \langle u, v \rangle_V \forall v \in V$. Let $L^* : W \rightarrow V$ be $L^*(w) = u$. Thus,

$\langle w, L(v) \rangle_W = \langle L^*(w), v \rangle_V \forall v \in V, w \in W$. Show linearity and uniqueness. Prop 3.7.12 $(S + T)^* = S^* + T^*$ and $(\alpha T)^* = \bar{\alpha} T^*$ OrthComplement of $S \subset V$ is the set $S^\perp = \{y \in V | \langle x, y \rangle = 0, \forall x \in S\}$. Note S^\perp is a subset of V . If W is finite dim. subspace of V , then $V = W \oplus W^\perp$. Fund. Subspaces: $\mathcal{R}(L)^\perp = \mathcal{N}(L^*)$ and $\mathcal{N}(L)^\perp = \mathcal{R}(L^*)$ COR: Let V, W be finite, $L : V \rightarrow W$ $V = \mathcal{N}(L) \oplus \mathcal{R}(L^*)$ and $W = \mathcal{R}(L) \oplus \mathcal{N}(L^*)$. Least squares: $\hat{\mathbf{x}} = (A^H A)^{-1} A^H \mathbf{b}$ is unique minimizer. Semi-Spectral mapping: If λ_i are the eigenvalues of a semisimple matrix $A \in M_n(\mathbb{F})$, and if $f(x)$ is any polynomial, then $\{f(\lambda_i)\}$ are the eigenvalues of $f(A)$. ExcerCh3: For reals,

$\langle \mathbf{x}, \mathbf{y} \rangle = \frac{1}{2} (\|\mathbf{x} + \mathbf{y}\|^2 - \|\mathbf{x} - \mathbf{y}\|^2)$ and $\|\mathbf{x}\|^2 + \|\mathbf{y}\|^2 = \frac{1}{2} (\|\mathbf{x} + \mathbf{y}\|^2 + \|\mathbf{x} - \mathbf{y}\|^2)$ CH4: Eigenvalues and eigenvectors depend only on the linear transformation and not the choice of basis of it. Let C_{TS} be the transition matrix, $A \in S, B \in T$ s.t. $[x]_T = C[x]_S$ and $B = C A C^{-1}$. Thus, $B[x]_T = C A C^{-1} C[x]_S = C \lambda[x]_S = \lambda C[x]_S = \lambda [x]_T$ THM: Following are equivalent, (i) λ is an eigenvalue (ii) There is a nonzero x such that $(\lambda I - A)x = 0$ (iii) $\Sigma_\lambda(A) \neq \{0\}$ (iv) $\lambda I - A$ is singular, thus $\det = 0$. Prop: If A, B are similar, that is $A = P B P^{-1}$. They are the same operator, and (i) have the same characistic poly, eigenvalues, the eigenbases are isomorphic. Invariant: A subspace is invariant if for $L : V \rightarrow V$, $L(W) \subset W$. A is simple if eigenvalues are distinct, semi-simple if eigenbasis spans A . Diagonalizable iff semisimple. $A = P D P^{-1}$. P is eigenvectors, D is eigenvalues. Fibonacci Numbers, Make the matrix, calculate eigens, note 300th number is $v_{301} = A v_{300} = A^{300} v_0 = P^{-1} D^{300} P v_0$ where v_0 is starting vector of sequence, of course, round to nearest integer. Power Method: pick vector, multiply by A , normalize, iterate. Implies dominant eigenvector and value if semi-simple. Rayleigh quotient does the same thing, but faster. $\frac{\langle \mathbf{x}, A\mathbf{x} \rangle}{\|\mathbf{x}\|^2}$ Implies λ and eigenvector, which converge to "some" eigenvalue and vector. Orthonormally similar if $B = U^H A U$, U is an orthonormal iff $\langle U\mathbf{x}, U\mathbf{y} \rangle = \langle \mathbf{x}, \mathbf{y} \rangle = x^H y$. Hermitian matrices are linear operators that preserve length and angle on different bases. Lem: If A is Hermitian, ortho similar to B , then B is Hermitian. Schur's Lemma: Every $n \times n$ matrix, A , is orthonormally similar to an upper triangular matrix. Proved by induction. Spet Thm I: Every Hermitian matrix A is orthonormally similar to a real diagonal matrix. Proof: By Schur's, A is ortho to an upper triangular, T . Since A is hermitian, so is T , and $T^H = T$ thus all eigenvalues (diagonals) are real. Normal Matrix: Spet Thm holds for all Normals. Normal when $A^H = A A^H$. Spet Thm II: A matrix A is normal iff it is orthonormally similar to a diagonal matrix, equivalently, if it has an orthonormal eigenbasis. Proof: by Shur's T is uppertriangular Other

direction, just multiply out. Positive Definite: Has positive eigen values, there exists a unique lower triangular matrix L , with real and strictly positive diagonal elements such that $M = LL^*$, that's cholesky decomp. It is invertible, and it's inverse is positive definite. Sum and product of semi definites are semi definite. $Q^T M Q$ is positive semi definite. Determinant is bounded by product of diagonal elements. $A = S^H S$, and if A is definite, S is nonsingular. SVD: if A is of rank r , \exists orthonormal U, V and real valued $\Sigma = \text{diag}(\sigma_1, \dots, \sigma_r, \dots, 0, 0, \dots)$ s.t. $A = U \Sigma V^H$ where σ_i is positive real valued. Σ is unique. To calculate, $\Sigma = \sqrt{\lambda_i}$ of $A^H A$. V is construction of eigenvectors of $A^H A$, with unfilled spots filled by gramscmidt (if eigenbasis doesn't exist). U is determined by $u_i = \frac{1}{\sigma_i} A v_i$ where u_i, v_i are columns of U, V respectively. Compact form is where we drop all zeros of sigma, and fit U, V accordingly. Moore-Penrose: $A^\dagger = V_1 \Sigma_1^{-1} U_1^H$ of compact, or $V \Sigma^\dagger U^*$ where Σ^\dagger is reciprocal of non-zeros on diagonal, transpose. Schmidt-Eckart-Young: for A of rank r , and each $s < r$, $\sigma_{s+1} = \inf_{\text{rank}(B)=s} \|A - B\|_2$ with minimizer $B^o = \sum_{i=1}^s \sigma_i u_i v_i^H$ where each σ_i is the singular value of A , with corresponding u, v columns of U, V in the compact form of SVD. Schmidt-Eckart-Young-Mirsky: $(\sum_{j=s+1}^r \sigma_j^2)^{1/2} = \inf_{\text{rank}(B)=s} \|A - B\|_F$ Cor: $A = U \Sigma V^H$ is the SVD where A has rank $r > s$ then for any $m \times n$ Δ such that $A + \Delta$ has rank s we have: $\|\Delta\|_2 \geq \sigma_{s+1}$ and $\|\Delta\|_F \geq (\sum_{k=s+1}^r \sigma_k^2)^{1/2}$ and equality holds if $\Delta = -\sum_{i=s+1}^r \sigma_i u_i v_i^H$. The infimum of $\|\Delta\|$ rank($I - A\Delta$) $< m$ is i/σ_1 with minimizer $\Delta^* = \frac{1}{\sigma_1} v_1 u_1^H$ Small gains: if $A \in M_n$, then $I - A\Delta$ is nonsingular, provided that $\|A\|_2 \|\Delta\|_2 < 1$ For $A \in M_{m \times n}(\mathbb{R})$: (i) $\|A\|_2 = \sigma_1$ (largest singular value) (ii) if A is invertible, then $\|A^{-1}\|_2 = \frac{1}{\sigma_n}$ (iii) $\|A^H\|_2^2 = \|A^T\|_2^2 = \|A^H A\|_2 = \|A\|_2^2$ (iv) if U, V are orthonormal then $\|UAV\|_2 = \|A\|_2$ (v) $\|UAV\|_F = \|A\|_F$ (vi) $\|A\|_F = (\sigma_1^2 + \dots + \sigma_n^2)^{1/2}$. (vii) $|\det(A)| = \prod \sigma_i$. The SVD gives an orthonormal basis of four fundamental subspaces. first r columns of V are basis of $\mathcal{R}(A^H)$, the last $n-r$ columns of V span $\mathcal{N}(A)$, first r columns of U span $\mathcal{R}(A)$ last $m-r$ span $\mathcal{N}(A^H)$. Remark: Let V, W be normed linear spaces, if $L \in \mathcal{B}$ then the induced norm on L satisfies $\|Lx\|_W \leq \|L\|_{V,W} \|x\|_V$. Fun Fact: $0 \leq \langle x - \lambda y, x - \lambda y \rangle = \langle x, x \rangle - \langle \lambda y, x \rangle - \langle x, \lambda y \rangle + \langle \lambda y, \lambda y \rangle = \langle x, x \rangle - \bar{\lambda} \langle y, x \rangle - \lambda \langle x, y \rangle + \bar{\lambda} \lambda \langle y, y \rangle$. Polar Decomp: If $A \in M_{m \times n}(\mathbb{C})$ and $m \geq n$. Then there exists a Q with orthonormal columns and positive semidefinite P such that $A = PQ$. Because $A = U \Sigma V^H = UV^H V \Sigma V^H$. Let $Q = UV^H$ and $P = V \Sigma V^H$ Norms: 1-Norm max sum of columns. ∞ -Norm max sum of rows. 2-Norm is largest singular value. Semi-Simple: Diagonalizable, Eigenbasis exists, distinct eigenvalues iff simple. Semi Positive Definite: $\langle x, Ax \rangle \geq 0$, non-negative eigenvalues, chebysky decomp, sub-matrix positive, sort of kind of hermitian. Normal: Orthonormally similar to Diagonal matrix, $A^H A = A A^H$, Hermitians, skewes, orthonormals... **CHAPTER 5: DEF: Metric** (i) Positive Definiteness $d(x, y) \geq 0$ (iff $x = y$) (ii) Symmetry: $d(x, y) = d(y, x)$ (iii) Triangle Inequality: $d(x, y) \leq d(x, z) + d(z, y)$. **DEF: Open Ball** with center at x_0 radius $r > 0$ to be the set $B(x_0, r) = \{x \in X | d(x, x_0) < r\}$. **DEF: Neighborhood** of point x_0 is a subset $E \subset X$ if \exists an open ball $B(x_0, r)$ such that $B(x_0, r) \subset E$. Here x_0 is an **interior point**. And E° is the **set of interior points of E**. **DEF: Open Set** if $E \subset X$ and every point $x \in E$ is an interior point. **THRM 5.1.10:** If $x \in B(x_0, r)$ then $B(x, r - \varepsilon) \subset B(x_0, r)$ where $\varepsilon = d(x_0, x)$. **THRM 5.1.12:** The union of any collection of open sets is open, and the intersection of any finite collection of open sets is open. **THRM 5.1.14:** (i) E° is open (ii) If G is an open subset of E , then $G \subset E^\circ$. (iii) E is open iff $E = E^\circ$ (iv) E° is the union of all open sets contained in E . **DEF: Continuous(1)** Let (X, d) and (Y, ρ) be metric spaces. A function $f: X \rightarrow Y$ is continuous at a point $x_0 \in X$ if for all $\varepsilon > 0$, there exists $\delta > 0$ such that $\rho(f(x), f(x_0)) < \varepsilon$ whenever $d(x, x_0) < \delta$. The function f is continuous on a subset $E \subset X$ if it is continuous at all $x_0 \in E$. The set of continuous functions from X to Y is denoted $C(X; Y)$. **THRM 5.1.18 Continuous(2):** A function $f: X \rightarrow Y$ is continuous on X iff the preimage $f^{-1}(U)$ of every open set $U \subset Y$ is open in X . **COR/PROP 5.1.19/20:** Composition, sum, product, and scalar multiples of continuous functions are continuous. **DEF: Limit Point** is a point $p \in E$ of set $E \subset X$ if every neighborhood of p intersects $E \setminus \{p\}$. **DEF: Isolated Point** is a point $p \in E$ where $E \subset X$ and p is not a limit point. **DEF: Dense** We say that E is dense in X if every point in X is either in E or is a limit point of E . **DEF: Closed** We say that F is closed if it contains all of its limit points. **THRM 5.2.8:** If p is a limit point of $E \subset X$, then every neighborhood of p contains infinitely many points. **THRM 5.2.9:** A set $U \subset X$ is open iff its compliment U^c is closed. **COR 5.2.12:** Intersection of any collection of closed sets is closed, and the union of a finite collection of closed sets is closed. **COR 5.2.14:** A function $f: X \rightarrow Y$ is continuous iff for each closed set $F \subset Y$ the preimage $f^{-1}(F)$ is closed on X . **DEF 5.2.16:** The closure of E , denoted \bar{E} , is the set E together with its limit points. We define the **boundary** of E , denoted $\text{Bd } E$, as the closure minus the interior, that is, $\text{Bd } E = \bar{E} \setminus E^\circ$. **DEF Limit:** A function $f: X \rightarrow Y$ has a limit $y_0 \in Y$ at $x_0 \in X$ if for all $\varepsilon > 0$, there exists a $\delta > 0$ such that $\rho(f(x), y_0) < \varepsilon$ whenever $0 < d(x, x_0) < \delta$. We denote the limit as $\lim_{x \rightarrow x_0} f(x) = y_0$. **DEF Limit (sequence):** The limit of the sequence $(x_i)_{i=0}^\infty$ if for all $\varepsilon > 0$, there exists $N > 0$

such that $d(\mathbf{x}, \mathbf{x}_n) < \varepsilon$ whenever $n \geq N$. We write $\mathbf{x}_k \rightarrow \mathbf{x}$ or $\lim_{k \rightarrow \infty} \mathbf{x}_k = \mathbf{x}$, and say that the sequence converges to \mathbf{x} . **DEF Cluster Point:** We say that $\mathbf{x} \in X$ is a cluster point of the sequence $(\mathbf{x}_i)_{i=0}^\infty$ if for all $\varepsilon > 0$ and $N > 0$, there exists $n \geq N$ such that $d(\mathbf{x}, \mathbf{x}_n) < \varepsilon$. **PROP 5.2.30:** A convergent sequence has exactly one cluster point. In particular, if a sequence has a limit, it is unique. **THRM 5.2.32:** The function $f: X \rightarrow Y$ is continuous at a point $\mathbf{x}_0 \in X$ iff for each sequence $(\mathbf{x}_i)_{i=0}^\infty \subset X$ that converges to \mathbf{x}_0 , the sequence $(f(\mathbf{x}_i))_{i=0}^\infty$ converges to $f(\mathbf{x}_0) \in Y$. **DEF: Cauchy sequence** A sequence $(\mathbf{x}_i)_{i=0}^\infty$ in X is a Cauchy sequence if for all $\varepsilon > 0$ there exists an $N > 0$ such that $d(\mathbf{x}_m, \mathbf{x}_n) < \varepsilon$ whenever $m, n \geq N$. **PROP 5.3.4:** Any sequence that converges is a Cauchy sequence. **PROP 5.3.6:** Cauchy sequences are **bounded** (for all $\mathbf{x} \in M \in \mathbb{N}$ such that for all \mathbf{p} we have $d(\mathbf{x}, \mathbf{p}) < M$). **PROP 5.3.8:** Any Cauchy sequence which has a convergent subsequence is convergent. **DEF: Complete** A metric space (X, d) is complete if every Cauchy sequence converges. **THRM 5.3.13:** The fields \mathbb{R} and \mathbb{C} are complete with respect to the usual metric $d(z, w) = |z - w|$. **THRM 5.3.14:** If $\{(X_i, d_i)\}_{i=1}^n$ is a finite collection of complete metric spaces, then the Cartesian product $X = X_1 \times X_2 \times \cdots \times X_n$ is complete when endowed with the p-metric for $1 \leq p \leq \infty$. **THRM 5.3.15:** Let $\mathbb{F} = \mathbb{C}$ or $\mathbb{F} = \mathbb{R}$. For every $n \in \mathbb{N}$ and $p \in [1, \infty]$, the linear space \mathbb{F}^n with the norm $\|\cdot\|_p$ is complete. **DEF: Uniform Continuity** A function $f: X \rightarrow Y$ is uniformly continuous on $E \subset X$ if for all $\varepsilon > 0$ there exists $\delta > 0$ such that $\rho(f(\mathbf{x}), f(\mathbf{y})) < \varepsilon$ whenever $\mathbf{x}, \mathbf{y} \in E$ and $d(\mathbf{x}, \mathbf{y}) < \delta$. **PROP 5.3.21:** If $f: X \rightarrow Y$ is a bounded linear transformation of normed linear spaces, then f is uniformly continuous. **THRM 5.3.23:** Let (X, d) and (Y, ρ) be metric spaces and $f: X \rightarrow Y$ be uniformly continuous. If $(\mathbf{x}_i)_{i=0}^\infty$ is a Cauchy sequence, then so is $(f(\mathbf{x}_k))_{k=0}^\infty$. **COR 5.3.24:** Every finite-dimensional normed linear space is complete with respect to the metric induced by the norm. **LEM 5.3.25:** If Y is a dense subspace of a normed linear space Z such that every Cauchy sequence in Y converges in Z , then Z is complete. **DEF: Open Cover** A collection $\{G_\alpha\}_{\alpha \in J}$ of open sets is an open cover of the set E if $E \subset \bigcup_{\alpha \in J} G_\alpha$. **DEF: Compact** A set E is compact if every open cover has a finite subcover, that is, for every open cover $\{G_\alpha\}_{\alpha \in J}$ there exists a finite subcollection $\{G_\alpha\}_{\alpha \in J'}$, where $J' \subset J$ is a finite subset, such that $E \subset \bigcup_{\alpha \in J'} G_\alpha$. **PROP 5.4.3:** A closed subset of a compact set is compact. **LEM 5.4.4:** Every infinite subset of a compact set K contains a limit point of K . **DEF: Bounded** We say that a subset K of a metric space X is bounded if there exists an $\mathbf{x} \in X$ and an $M > 0$ such that $K \subset B(\mathbf{x}, M)$. **THRM 5.4.6:** A compact subset of a metric space is closed and bounded. **Heine-Borel Theorem:** Consider \mathbb{R}^n with the usual (Euclidean) metric. If a subset is closed and bounded, then it is compact. **PROP 5.4.9:** The continuous image of a compact set is compact; that is, if $f: X \rightarrow Y$ is continuous and $K \subset X$ is compact, then $f(K) \subset Y$ is compact. **Extreme-Valued Theorem:** Let (X, d) be a metric space. If $f: X \rightarrow \mathbb{R}$ is continuous and $K \subset X$ is a nonempty compact set, the $f(K)$ contains its infimum and supremum. **THRM 5.4.12:** Let (X, d) and (Y, ρ) be metric spaces and $K \subset X$ be compact. If $f: K \rightarrow Y$ is continuous, then f is uniformly continuous on K . **THRM 5.4.14:** The following are equivalent: (i) X is compact (ii) Every collection \mathcal{C} of closed sets in X with the finite intersection property has a non-empty intersection (iii) X has the Bolzano-Weierstrass property (every infinite sequence $(\mathbf{x}_k)_{k=0}^\infty \subset X$ has at least one cluster point) (iv) X is sequentially compact (every sequence $(\mathbf{x}_k)_{k=0}^\infty \subset X$ has a convergent subsequence) (v) X is totally bounded (for all $\varepsilon > 0$ the cover $\mathcal{C} = \{B(\mathbf{x}, \varepsilon)\}_{\mathbf{x} \in X}$ has a finite subcover) and every open cover has a positive Lebesgue number (which depends on the cover). **Generalized Heine-Borel Theorem:** A metric space X is compact iff it is complete and totally bounded. **DEF: Pointwise Convergence** For any sequence of functions $(f_n)_{n=0}^\infty$, we can evaluate all the functions at a single point in the domain, which gives the sequence $(f_n(x))_{n=0}^\infty$. If for every x , the sequence converges, then we can define a new function given by $f(x) = \lim_{n \rightarrow \infty} f_n(x)$. In this case the sequence converges pointwise to $f(x)$. **DEF: Uniform Convergence** Let $(f_k)_{k=0}^\infty$ be a sequence of bounded functions from some domain X to a normed space Y . If $(f_k)_{k=0}^\infty$ converges to f in the L^∞ norm, then we say that the sequence $(f_k)_{k=0}^\infty$ converges uniformly to f . **PROP 5.5.5:** Uniform convergence implies pointwise convergence. **DEF: Banach Space** A complete normed linear space is called a Banach Space. **THRM 5.5.8:** For any $a < b \in \mathbb{R}$ the space $(C([a, b]; \mathbb{F}), \|\cdot\|_{L^\infty})$ is a Banach space. **COR 5.5.9:** If a sequence of continuous functions converges uniformly to a function f , then f is also continuous. **DEF: Convergence (Banach)** Let $(X, \|\cdot\|_X)$ be a Banach space and consider the sequence $(\mathbf{x}_k)_{k=0}^\infty \subset X$. We say that the series $\sum_{k=0}^\infty \mathbf{x}_k$ converges in X if the sequence $(\mathbf{s}_k)_{k=0}^\infty$ of partial sums defined by $\mathbf{s}_n = \sum_{k=1}^n \mathbf{x}_k$ converges in X ; otherwise we say that the series diverges. **DEF: Absolute Convergence (Banach)** Let $(\mathbf{x}_k)_{k=0}^\infty$ be a sequence in the Banach space $(X, \|\cdot\|_X)$. The series $\sum_{k=0}^\infty \mathbf{x}_k$ is said to absolutely converge if the series $\sum_{k=0}^\infty \|\mathbf{x}_k\|_X$ converges in \mathbb{R} . **PROP 5.5.13:** Let $(\mathbf{x}_k)_{k=0}^\infty$ be a sequence in the Banach space $(X, \|\cdot\|_X)$. If the series $\sum_{k=0}^\infty \mathbf{x}_k$ converges absolutely then it converges in X . **EX 5.5.20:** Let $(X, \|\cdot\|_X)$ be a Banachspace. Let $T \in \mathcal{B}(X)$ be a bounded operator with $\|T\| < 1$. The **Neumann Series** of A is the sum $\sum_{k=0}^\infty A^k$. If $\|\cdot\|$ is the operator norm, then:

$\sum_{k=0}^{\infty} \|A^k\| \leq \sum_{k=0}^{\infty} \|A\|^k = \frac{1}{1-\|A\|} < \infty$. **PROP 5.5.21:** Let $(X, \|\cdot\|_X)$ be a Banach space. If $A \in \mathcal{B}(X)$ satisfies $\|A\| < 1$ then $I - A$ is invertible. Moreover, we have that $(I - A)^{-1} = \sum_{k=0}^{\infty} A^k$, and thus $\|(I - A)^{-1}\| \leq (1 - \|A\|)^{-1}$. **Topological Properties** preserved under homeomorphism: open sets, continuity, compactness, convergence, Connectedness **Not Topological:** Completeness, Cauchy (unless using topologically equivalent norms) **DEF: Bounded Functions** Let $(X, \|\cdot\|_X)$ be a Banach space. Write $L^\infty([a, b]; X)$ to denote the set of bounded functions $f: [a, b] \rightarrow X$ equipped with the sup norm $\|f\|_\infty = \sup_{t \in [a, b]} \|f(t)\|$. **THRM 5.7.2:** $L^\infty([a, b]; X)$ is a Banach space. **DEF: Step Function** Let $(X, \|\cdot\|_X)$ be a Banach space. A map $f: [a, b] \rightarrow X$ is a step function if there is a (finite) partition $a = t_0 < t_1 < \dots < t_{N-1} < t_N = b$ such that we may write f in the form: $f(t) = \sum_{i=1}^{N-1} \mathbf{x}_i \mathbf{1}_{[t_{i-1}, t_i]} + \mathbf{x}_N \mathbf{1}_{[t_{N-1}, t_N]}$, where each $\mathbf{x}_i \in X$ and $\mathbf{1}_E$ is the indicator function $\mathbf{1}_E(t) = 1$ if $t \in E$, 0 if $t \notin E$. **Continuous Linear Extension Theorem:** Let $(Z, \|\cdot\|_Z)$ be a normed linear space, $(X, \|\cdot\|_X)$ a Banach space, and $S \subset Z$ a dense subspace of Z . If $T: S \rightarrow X$ is a bounded linear transformation, then T has a unique linear extension to $\bar{T} \in \mathcal{B}(Z; X)$ satisfying $\|\bar{T}\| = \|T\|$. **DEF: Integral Properties** If $(X, \|\cdot\|)$ is a Banach space, if $f \in \mathcal{S}([a, b]; X) \subset L^\infty([a, b]; X)$ and $\alpha, \beta, \gamma \in [a, b]$, with $\alpha < \gamma < \beta$, then (i) $\| \int_\alpha^\beta f(t) dt \| \leq \int_\alpha^\beta \|f(t)\| dt \leq (b - a) \sup_{t \in [a, b]} \|f(t)\|$ (ii) Restricting f to $[\alpha, \beta]$ defines a function that we also denote by $f \in \mathcal{S}([a, b]; X)$. We have $\int_\alpha^\beta f(t) \mathbf{1}_{[\alpha, \beta]} dt = \int_\alpha^\beta f(t) dt$ (iii) $\int_\alpha^\beta f(t) dt = \int_\alpha^\gamma f(t) dt + \int_\gamma^\beta f(t) dt$. (iv) $F(t) = \int_a^t f(s) ds$ is continuous on $[a, b]$.

Ch.6 The directional derivative of f at x with respect to v is the limit $\lim_{t \rightarrow 0} \frac{f(x+tv) - f(x)}{t}$ this limit is often denoted $D_v f(x)$. **partial derivatives** the i th partial derivative of f at the point x is given by the limit $D_i f(x) = \frac{f(x+h\mathbf{e}_i) - f(x)}{h}$. **Frechet derivative*:** Let $(X, \|\cdot\|_X)$ and $(Y, \|\cdot\|_Y)$ be Banach spaces and let $U \subset X$ be an open set. A map f is differentiable at $x \in U$ if there exists a bounded linear transformation $Df(x): X \rightarrow Y$ s. t. $\lim_{h \rightarrow 0} \frac{\|f(x+h) - f(x) - Df(x)h\|_Y}{\|h\|_X} = 0$. X and Y are Banach spaces, $U \subset X$ be an open set, and f be differentiable on U . If Df given by $x \rightarrow Df(x)$ is also continuous, we say that f is continuously differentiable on U . if f is differentiable on U , then f is locally **lipshitz**, that is $\forall x_0 \in U \exists B(x_0, \delta) \subset U$ and $L > 0$ s.t. $\|f(x) - f(x_0)\|_Y \leq L\|x - x_0\|_X$ whenever $\|x - x_0\|_X < \delta$.

Mean Value Theorem: Let $(X, \|\cdot\|_X)$ be a Banach space, $U \subset X$ be an open set, and $f: U \rightarrow \mathbb{R}$ be differentiable on U . If for $\mathbf{x}, \mathbf{x} \in U$, the entire line segment $\ell(\mathbf{x}, \mathbf{x}) := \{(1-t)\mathbf{x} + t\mathbf{x} \in [0, 1]\}$ is also in U , then there exists $\mathbf{c} \in \ell(\mathbf{x}, \mathbf{x})$ such that: $f(\mathbf{y}) - f(\mathbf{x}) = Df(\mathbf{c})(\mathbf{y} - \mathbf{x})$. **Fundamental Theorem of Calculus:** Let $(X, \|\cdot\|_X)$ be a Banach space: (i) If $f \in C([a, b]; X)$, then for all $t \in (a, b)$ we have that: $\frac{d}{dt} \int_a^t f(s) ds = f(t)$ (ii) If $F: [a, b] \rightarrow X$ is continuously differentiable on (a, b) and $DF(t)$ extends to a continuous function on $[a, b]$, then: $\int_a^b DF(s) ds = F(b) - F(a)$. **Integral Mean Value Theorem:** Let $(X, \|\cdot\|_X)$ and $(Y, \|\cdot\|_Y)$ be Banach spaces, $U \subset X$ be an open set, and $f: U \rightarrow Y$ be continuously differentiable on U . If the line segment $\ell(\mathbf{x}, \mathbf{x}) = \{t\mathbf{x} + (1-t)\mathbf{x} | t \in [0, 1]\}$ is contained in U , then: $f(\mathbf{x}) - f(\mathbf{x}) = \int_0^1 Df(t\mathbf{y} + (1-t)\mathbf{x})(\mathbf{x} - \mathbf{x}) dt$, alternatively if we let $\mathbf{y} = \mathbf{x} + \mathbf{h}$, then $f(\mathbf{x} + \mathbf{h}) - f(\mathbf{x}) = \int_0^1 Df(\mathbf{x} + t\mathbf{h})\mathbf{h} dt$ or $\|f(\mathbf{x}) - f(\mathbf{x})\|_Y \leq \sup_{\mathbf{c} \in \ell(\mathbf{x}, \mathbf{x})} \|Df(\mathbf{c})\|_{X,Y} \|\mathbf{x} - \mathbf{x}\|_X$ **Change of Variable Formula:** Let $(X, \|\cdot\|_X)$ be a Banach space and $f \in C([a, b]; X)$. If $g: [c, d] \rightarrow [a, b]$ is continuous and g' is continuous on (c, d) and can be continuously extended to $[c, d]$, then $\int_c^d f(g(s))g'(s)ds = \int_{g(c)}^{g(d)} f(\tau)d\tau$

Lemma*: Given $f: X \rightarrow Y$ and linear $L: X \rightarrow Y$, to prove that f is differentiable at x with derivative L is equivalent to proving that for every $\epsilon > 0$ there is $\|\xi\| < \delta$ we have $\|f(x + \xi) - f(x) - L\xi\|_Y \leq \epsilon\|\xi\|_X$. **Linearity:** assume X and Y are Banach spaces, that U is an open neighborhood in X , and $f: U \rightarrow Y$ and $g: U \rightarrow Y$. If f and g are differentiable on U and $a, b \in \mathbb{R}$, then $af + bg$ is also differentiable on U , and $D(af(x) + bg(x)) = aDf(x) + bDg(x)$. **Product Rule:** Let X be Banach space, that U is an open neighborhood of X , and that $f: U \rightarrow \mathbb{R}$ and $g: U \rightarrow \mathbb{R}$ if f and g are differentiable on U , then the product map fg is also differentiable on U and $D(f(x)g(x)) = g(x)Df(x) + f(x)Dg(x)$ for each $x \in U$. Rules of differentiation: (i) $u(x), v(x)$ are differentiable from \mathbb{R}^n to \mathbb{R}^m and $f: \mathbb{R}^m \rightarrow \mathbb{R}$ then $Df(x) = u(x)^T Du(x) + v(x)^T Dv(x)$. (ii) if $g: \mathbb{R}^n \rightarrow \mathbb{R}$ s.t. $g(x) = x^T Ax$ then $Dg(x) = x^T(A + A^T)$. **Chain Rule:** Assume X, Y , and Z are Banach, U and V are open neighborhoods of X and Y respectively, and $f: U \rightarrow V$ and $g: V \rightarrow Z$ with $f(U) \subset V$ if f is differentiable on U and g is differentiable on V , then the composite map $h = f \circ g$ is also differentiable on U and $Dh(x) = Dg(f(x))Df(x)$, for each $x \in U$

Sec6.5 Let $U \subset \mathbb{R}^n$ be open. If $f: U \rightarrow \mathbb{R}^m$ is twice continuously differentiable on U , then $D^2 f(x)(x_1, x_2) = D^2 f(x)(x_2, x_1)$. Or $\frac{\partial^2 f_{ij}}{\partial x_i \partial x_j} = \frac{\partial^2 f_{ji}}{\partial x_j \partial x_i}$ Taylor's Theorem: For X, Y be Banach. $U \subset X$ an open set, and $f: U \rightarrow Y$ be k times differentiable. If $x \in U$ and $h \in X$ are such that the line $l(x, x + h)$ is contained in U , then

$f(x + h) = f(x) + Df(x)h + \frac{D^2 f(x)h^2}{2!} + R_k$. Example: for $f(x, y) = e^{x+y}$ at $(0, 0)$ in direction of (h_1, h_2) . Note, $f(0, 0) = 1$. $D_h f(x) = \nabla f(0, 0) \cdot h = h_1 D_1(0, 0) + h_2 D_2(0, 0) = h_1 + h_2$.