

MATH 2242 (Calculus IV) Course Outline — Vector Calculus (Marsden)

1.5 n -Dimensional Euclidean Space

- $\mathbb{R}, \mathbb{R}^2, \mathbb{R}^3, \mathbb{R}^n$
- Addition
 - $\langle x_1, x_2, \dots, x_n \rangle + \langle y_1, y_2, \dots, y_n \rangle = \langle x_1 + y_1, x_2 + y_2, \dots, x_n + y_n \rangle$
- Scalar multiplication
 - $\alpha \langle x_1, x_2, \dots, x_n \rangle = \langle \alpha x_1, \alpha x_2, \dots, \alpha x_n \rangle$
- Inner/Dot Product
 - $\langle x_1, x_2, \dots, x_n \rangle \cdot \langle y_1, y_2, \dots, y_n \rangle = \sum_{i=1}^n x_i y_i$
- Standard basis vectors
 - $\mathbf{e}_1 = \langle 1, 0, \dots, 0 \rangle, \mathbf{e}_2 = \langle 0, 1, \dots, 0 \rangle, \dots, \mathbf{e}_n = \langle 0, 0, \dots, 1 \rangle$
- Theorems
 - $(\alpha \mathbf{x} + \beta \mathbf{y}) \cdot \mathbf{z} = \alpha(\mathbf{x} \cdot \mathbf{z}) + \beta(\mathbf{y} \cdot \mathbf{z})$
 - Prove the above theorem.
 - $\mathbf{x} \cdot \mathbf{y} = \mathbf{y} \cdot \mathbf{x}$
 - $\mathbf{x} \cdot \mathbf{x} \geq 0$
 - $\mathbf{x} \cdot \mathbf{x} = 0$ if and only if $\mathbf{x} = \mathbf{0}$
 - $|\mathbf{x} \cdot \mathbf{y}| \leq \|\mathbf{x}\| \|\mathbf{y}\|$ (the Cauchy-Schwarz inequality)
 - Prove the Cauchy-Schwarz inequality.
 - $\|\mathbf{x} + \mathbf{y}\| \leq \|\mathbf{x}\| + \|\mathbf{y}\|$ (the triangle inequality)
 - Prove the triangle inequality.
- Matrices
 - $A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix}$
 - Addition $A + B$
 - Scalar Multiplication αA

■ Transposition A^T

• Vectors as Matrices

■ $\mathbf{a} = \langle a_1, a_2, \dots, a_n \rangle = \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix}$

■ $\mathbf{a}^T = [a_1 \ a_2 \ \cdots \ a_n]$

• Matrix Multiplication

- If A has m rows and B has n columns, then $M = AB$ is an $m \times n$ matrix.
- Coordinate ij of $M = AB$ is given by $m_{ij} = \mathbf{a}_i \cdot \mathbf{b}_j$ where \mathbf{a}_i^T is the i th row of A and \mathbf{b}_j is the j th column of B .
- (Example 4) Compute AB and BA for

$$A = \begin{bmatrix} 1 & 0 & 3 \\ 2 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}$$

$$B = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 1 \end{bmatrix}$$

- (Example 5) Compute AB for

$$A = \begin{bmatrix} 2 & 0 & 1 \\ 1 & 1 & 2 \end{bmatrix}$$

$$B = \begin{bmatrix} 1 & 0 & 2 \\ 0 & 2 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

• Matrices as Linear Transformations

- An $m \times n$ matrix A gives a function from \mathbb{R}^n to \mathbb{R}^m : $\mathbf{x} \mapsto A\mathbf{x}$
- This linear transformation satisfies $A(\alpha\mathbf{x} + \beta\mathbf{y}) = \alpha A\mathbf{x} + \beta A\mathbf{y}$

- (Example 7) Express $A\mathbf{x}$ where $x = \langle x_1, x_2, x_3 \rangle$ and $A = \begin{bmatrix} 1 & 0 & 3 \\ -1 & 0 & 1 \\ 2 & 1 & 2 \\ -1 & 2 & 2 \end{bmatrix}$. Then

compute where the points $(3, -2, 1)$, $(1, 0, 1)$, $(-1, 1, 0)$, and $(-3, 3, 0)$ in \mathbb{R}^3 get mapped to in \mathbb{R}^4

- Identity and Inverse

- $I_n = \begin{bmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 1 \end{bmatrix}$

- If $AA^{-1} = A^{-1}A = I_n$, then A is invertible and A^{-1} is its inverse.

- Determinant

- Let A_i be the submatrix of A with the first column and i th row removed. Then $\det(A) = \sum_{i=1}^n (-1)^{i+1} a_{1i} \det(A_i)$

- HW: 1-18, 21-24

2.3 Differentiation

- Functions $\mathbb{R}^n \rightarrow \mathbb{R}^m$

- $\mathbf{f} : \mathbb{R}^n \rightarrow \mathbb{R}^m$

- $\mathbf{f}(\mathbf{x}) = \langle f_1(\mathbf{x}), \dots, f_m(\mathbf{x}) \rangle$ where $f_i : \mathbb{R}^n \rightarrow \mathbb{R}$

- Partial Derivative Matrix

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

- We say \mathbf{f} is differentiable at \mathbf{x} if $\mathbf{f}(\mathbf{x} + \mathbf{h}) \approx \mathbf{f}(\mathbf{x}) + [\mathbf{Df}(\mathbf{x})]\mathbf{h}$ for all \mathbf{h} near $\mathbf{0}$.

- (Example) Let $\mathbf{f} : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ be defined by $\mathbf{f}(x, y) = \langle x^2 + y^2, xy \rangle$, and let $\mathbf{T} = \mathbf{Df}(1, 0)$. Compute $\mathbf{f}(1.1, -0.1)$ and $\mathbf{f}(1, 0) + \mathbf{T}\langle 0.1, -0.1 \rangle$.

- If each $\frac{\partial f_i}{\partial x_j} : \mathbb{R}^n \rightarrow \mathbb{R}$ is a continuous function near \mathbf{x} , then we say \mathbf{f} is strongly differentiable at \mathbf{x} .

- Gradient

- If $f : \mathbb{R}^n \rightarrow \mathbb{R}$, then the gradient vector function $\nabla f : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is defined by $\nabla f(\mathbf{x}) = (\mathbf{Df}(\mathbf{x}))^T = \langle \frac{\partial f}{\partial x_1}(\mathbf{x}), \dots, \frac{\partial f}{\partial x_n}(\mathbf{x}) \rangle$

- $[\mathbf{Df}(\mathbf{x})]\mathbf{h} = \nabla f(\mathbf{x}) \cdot \mathbf{h}$

- HW: 1-21

2.5 Properties of the Derivative

- Sum/Product/Quotient Rules
 - $D[\alpha f] = \alpha Df$
 - $D[f + g] = Df + Dg$
 - $D[fg] = gDf + fDg$
 - $D\left[\frac{f}{g}\right] = \frac{gDf - fDg}{g^2}$
 - Sketch proofs for strongly differentiable f, g .
- Chain Rule
 - $D[f \circ g] = [Df](g)Dg$
 - (Example) Find the rate of change of $f(x, y) = x^2 + y^2$ along the path $\mathbf{c}(t) = \langle t^2, t \rangle$ when $t = 1$.
 - (Example 2) Verify the Chain Rule for $f(u, v, w) = u^2 + v^2 - w$ and $\mathbf{g}(x, y, z) = \langle x^2y, y^2, e^{-xz} \rangle$.
 - (Example 3) Compute $D[f \circ g](1, 1)$ where $\mathbf{f}(u, v) = \langle u + v, u, v^2 \rangle$ and $\mathbf{g}(x, y) = \langle x^2 + 1, y^2 \rangle$.
- HW: 6-13, 15-16

3.2 Taylor's Theorem

- First-Order Taylor Formula
 - $f(\mathbf{x} + \mathbf{h}) \approx f(\mathbf{x}) + [Df(\mathbf{x})]\mathbf{h}$
 - Alternate form: $f(\mathbf{x} + \mathbf{h}) \approx f(\mathbf{x}) + \sum_{i=1}^n \frac{\partial f}{\partial x_i}(\mathbf{x})h_i$
- Second-Order Taylor Formula
 - $f(\mathbf{x} + \mathbf{h}) \approx f(\mathbf{x}) + \sum_{i=1}^n \frac{\partial f}{\partial x_i}(\mathbf{x})h_i + \frac{1}{2} \sum_{i,j=1}^n \frac{\partial^2 f}{\partial x_i \partial x_j}(\mathbf{x})h_i h_j$
 - (Example 3) Find linear and quadratic functions of x, y which approximate $f(x, y) = \sin(xy)$ near the point $(1, \pi/2)$.
- HW: 1-12

Remaining Topics

- 4.1 Acceleration and Newton's Second Law
- 4.2 Arc Length
- 4.3 Vector Fields
- 4.4 Divergence and Curl
- 5.3 The Double Integral Over More General Regions
- 5.4 Changing the Order of Integration
- 5.5 The Triple Integral
- 6.1 The Geometry of Maps from \mathbb{R}^2 to \mathbb{R}^2
- 6.2 The Change of Variables Theorem
- 7.1 The Path Integral
- 7.2 Line Integrals
- 7.3 Parametrized Surfaces
- 7.4 Area of a Surface
- 7.5 Integrals of Scalar Functions Over Surfaces
- 7.6 Surface Integrals of Vector Fields
- 8.1 Green's Theorem
- 8.2 Stokes' Theorem
- 8.3 Conservative Fields
- 8.4 Gauss' Theorem