LECTURE 2: REVIEW

Disclaimer. As we have a textbook, this lecture note is for guidance and supplement only. It should not be relied on when preparing for exams.

In this lecture we review the prerequisites: Multivariable calculus, linear algebra, and basics of differential equations. We will not be able to cover everything that will be needed through the semester. We will only cover the basics, and leave some more sophisticated topics such as inverse/implicit function theorem for later lectures to explain.

Warning. In this lecture note we omit many technical assumptions such as differentiability/integrability of functions when stating formulas and theorems, because in 348 they are almost always satisfied. However please do not quote the statements here for your other courses.

TABLE OF CONTENTS

LECTURE 2: REVIEW 1
1. Multivariable calculus
1.1. The Euclidean space \mathbb{R}^n
1.2. Operations on \mathbb{R}^n 2
1.3. Functions from \mathbb{R}^m to \mathbb{R}^n
1.4. Taking derivatives of functions from \mathbb{R}^m to \mathbb{R}^n
2. Linear algebra
2.1. Operations on matrices
2.2. Linear dependence and independence
3. Differential equations
3.1. The simplest and second simplest differential equations
3.2. Existence and uniqueness of a general first order ODE

1. Multivariable calculus

Multivariable calculus studies functions from \mathbb{R}^m to \mathbb{R}^n with either m > 1 or n > 1. In 348 we focus on functions from \mathbb{R} to $\mathbb{R}^2 / \mathbb{R}^3$ (curves) and from \mathbb{R}^2 to \mathbb{R}^3 (surfaces). In the following we review concepts, theorems, and techniques that are important to us.

1.1. The Euclidean space \mathbb{R}^n

• \mathbb{R}^n is the mathematical representation of a *n*-dimensional flat space through setting up *n* orthogonal "axes". Along each axes, we define a unit vector: $e_1, ..., e_n$. Then each point in the space is identified by an *n*-tuple $(x_1, ..., x_n)$.

It is important to realize that \mathbb{R}^n can be interpreted in two ways, in the context of describing a moving particle (along a curve, on a surface, etc.).

- \mathbb{R}^n as possible locations. At each moment, the location of the particle is identified with a point in \mathbb{R}^n : At time t, the particle is at $(x_1, ..., x_n)$.
- \mathbb{R}^n as possible velocities. To completely describe the motion of this particle, we also need to know its velocity. The velocity is also described by an *n*-tuple $(v_1, ..., v_n)$. Such *n*-tuples are called "vectors".

Thus at each point x in the "location" space \mathbb{R}^n , we "overlay" a "velocity" space whose mathematical representation is also \mathbb{R}^n . This overlay will later be called the "tangent space" of the location space at x. Such distinction seems a bit silly now, but will be very useful later when we "curve" the location space. The key idea of differential geometry is that when the location space is curved, the velocity space stays flat. To visualize, think of a 2-dimensional surface as the curved locations space, then the velocity space at a location is the tangent plane of the surface there.

1.2. Operations on \mathbb{R}^n

- When \mathbb{R}^n is interpreted as the overlaying "velocity" space, the most important concepts are
 - NORM. When $v = (v_1, ..., v_n)$ describes a velocity vector, its speed is given by its norm:

$$\|v\| := \sqrt{v_1^2 + \dots + v_n^2}.$$
 (1)

• INNER PRODUCT. The inner product of two vectors $u = (u_1, ..., u_n)$ and $v = (v_1, ..., v_n)$ is defined as

$$u \cdot v := u_1 v_1 + \dots + u_n v_n. \tag{2}$$

The importance of inner product comes from the following property:

PROPOSITION 1. Let θ be the angle between two vectors $u = (u_1, ..., u_n)$ and $v = (v_1, ..., v_n)$, then

$$\cos\theta = \frac{u \cdot v}{\|u\| \|v\|}.\tag{3}$$

Exercise 1. Obviously θ is not uniquely defined. What does this not matter?

In particular, $u \cdot v = 0 \iff u \perp v$, that is u is perpendicular (orthogonal) to v.

• CROSS PRODUCT/VECTOR PRODUCT. Let $u = (u_1, u_2, u_3), v = (v_1, v_2, v_3)$ be two vectors in \mathbb{R}^3 . Their cross product (also called "vector product") is defined as

$$u \times v := (u_2 v_3 - u_3 v_2, u_3 v_1 - u_1 v_3, u_1 v_2 - u_2 v_1).$$
(4)

Exercise 2. Prove the following.

i. $u \times v = v \times u;$

ii. Let $a, b \in \mathbb{R}$ be arbitrary. Then

$$(a u + b w) \times v = a u \times v + b w \times v.$$
⁽⁵⁾

iii. $u \times v = 0$ if and only if u and v are "parallel".

iv. $(u \times v) \cdot u = 0, (u \times v) \cdot v = 0.$

v. $(u \times v) \cdot w = (v \times w) \cdot u = (w \times u) \cdot v.$

Also note that

$$\|u \times v\| = \|u\| \|v\| \sin\theta \tag{6}$$

where θ is the angle between u, v.

- When \mathbb{R}^n is interpreted as the underlying "location" space, the most important concepts are
 - DISTANCE. Let $x = (x_1, ..., x_n)$ and $y = (y_1, ..., y_n)$. For now, the only distance we know is the distance along straight lines, given by the Pythagorean theorem:

$$d(x, y) = \sqrt{(x_1 - y_1)^2 + \dots + (x_n - y_n)^2}.$$
(7)

Remark 2. We see that d(x, y) = ||x - y||. However this will cease to be true when the location space is curved. One important theme in 348 is to generalize (7) for curved spaces.

- AREA. The foundation of the definition of area is the following agreement:
 - i. The unit square has area 1.
 - ii. The area of a union of disjoint regions is the sum of the areas of these regions.
 - iii. Moving a region around "rigidly" (that is not changing the distance of any pair of two points in the region) does not change its area.

From these one could develop the whole theory of integration in two-variable calculus.

Example 3. Let $x = (x_1, x_2, x_3), y = (y_1, y_2, y_3)$ in \mathbb{R}^3 . Then the area of the parallelogram $(0, 0, 0) - (x_1, x_2, x_3) - (x_1 + y_1, x_2 + y_2, x_3 + y_3) - (y_1, y_2, y_3) - (0, 0, 0)$ is given by

$$\|x \times y\| := \|(x_2 y_3 - x_3 y_2, x_3 y_1 - x_1 y_3, x_1 y_2 - x_2 y_1)\|$$
(8)

- VOLUME. The foundation of the definition of volume is the following agreement:
 - i. The unit cube has volume 1.
 - ii. The volume of a union of disjoint regions is the sum of the volumes of these regions.
 - iii. Moving a region around "rigidly" (that is not changing the distance of any pair of two points in the region) does not change its volume.

With these agreement one could develop the whole theory of integration in three-variable calculus.

Example 4. Let $x = (x_1, x_2, x_3), y = (y_1, y_2, y_3), z = (z_1, z_2, z_3)$ in \mathbb{R}^3 . Then the volume of the parallelopiped spanned by these three vectors is given by

$$V = |(x \times y) \cdot z|. \tag{9}$$

Exercise 3. Check your old textbook and explain what the sign of $(x \times y) \cdot z$ means.

1.3. Functions from \mathbb{R}^m to \mathbb{R}^n

Let $\Omega \subseteq \mathbb{R}^m$ be a region in \mathbb{R}^m . A function $f: \Omega \mapsto \mathbb{R}^n$ is a correspondence between points in Ω and points in \mathbb{R}^n satisfying the following: Each $x \in \Omega$ corresponds to exactly one $y \in \mathbb{R}^n$. One classifies functions into two categories: linear and nonlinear. In 348, we will see that linear functions are often considered in the context of "velocity" spaces.

• LINEAR FUNCTIONS. A function $f: \Omega \mapsto \mathbb{R}^n$ is linear if and only if for every $a, b \in \mathbb{R}$ and every $x, y \in \Omega$, one has

$$f(a x + b y) = a f(x) + b f(y).$$
(10)

Exercise 4. Prove that if f is linear then f(0) = 0.

÷ ÷ ÷

Exercise 5. Let $f: \mathbb{R} \mapsto \mathbb{R}$ be linear. Prove that there is a constant $c \in \mathbb{R}$ such that f(x) = cx.

NOTATION. We often denote a linear function by a capital letter such as T^{1} .

PROPOSITION 5. Let $T: \Omega \subseteq \mathbb{R}^m \mapsto \mathbb{R}^n$ be linear. Then there are m n numbers a_{11} , $a_{12}, ..., a_{1m}, a_{21}, ..., a_{n1}, ..., a_{nm} \in \mathbb{R}$, such that for every $v \in \Omega$, the following holds for w = T(v):

$$w_1 = a_{11}v_1 + \dots + a_{1m}v_m \tag{11}$$

$$w_n = a_{n1}v_1 + \dots + a_{nm}v_m.$$
(12)

In other words, we have T(v) = A v where A is the matrix $\begin{pmatrix} a_{11} & \cdots & a_{1m} \\ \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nm} \end{pmatrix}$. Exercise 6. Prove Proposition 5.

^{1.} First letter of "transformation".

- NONLINEAR FUNCTIONS. Any function that is not linear. In calculus, nonlinear functions are studies through their derivatives.
- BILINEAR FORMS. A bilinear form is a special type of nonlinear function: $B: \mathbb{R}^{2n} \mapsto \mathbb{R}$, satisfying for every $a, b \in \mathbb{R}$ and every $u, v, w \in \mathbb{R}^n$,

$$B(a u + b w, v) = a B(u, v) + b B(w, v),$$
(13)

$$B(u, a w + b v) = a B(u, w) + b B(u, v).$$
(14)

Exercise 7. Prove that the inner product: $f(u, v) := u \cdot v$ is a bilinear form.

Exercise 8. Prove or disprove: The area $A(u, v) := ||u \times v||$ is a bilinear form.

PROPOSITION 6. Let $B: \mathbb{R}^{2n} \to \mathbb{R}$ be a bilinear form. Then there are n^2 numbers $a_{11}, \ldots, a_{1n}, a_{21}, \ldots, a_{n1}, \ldots, a_{nn}$ such that

$$B(u,v) = \sum_{i=1}^{n} \sum_{j=1}^{n} u_i a_{ij} v_j.$$
 (15)

Exercise 9. Let $A := \begin{pmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nn} \end{pmatrix}$. Prove that $B(u, v) = u \cdot A v = (A u) \cdot v$ where \cdot denotes inner product.

1.4. Taking derivatives of functions from \mathbb{R}^m to \mathbb{R}^n

• DIFFERENTIABILITY.

DEFINITION 7. A function $f: \Omega \subseteq \mathbb{R}^m \mapsto \mathbb{R}^n$ is differentiable at $x \in \Omega$ if and only if the following holds: There is a linear function $T: \mathbb{R}^m \mapsto \mathbb{R}^n$ such that

$$\lim_{v \to 0} \frac{\|f(x+v) - f(x) - T(v)\|}{\|v\|} = 0.$$
(16)

The linear function T is called the "differential" of f at x and will be denoted Df(x).

Remark 8. Note that f is defined on Ω which belongs to the "location" space \mathbb{R}^m , and f takes value in another "location" space \mathbb{R}^n ; while T is defined on \mathbb{R}^m which is the "velocity" overlay space on top of Ω , centered at x, and takes value in another "velocity space" centered at f(x).

By Proposition 5 there is a matrix A such that T(v) = A v for all $v \in \mathbb{R}^m$. The numbers a_{ij} in the matrix A is given by the partial derivatives of f at x:

DEFINITION 9. Let $f: \Omega \subseteq \mathbb{R}^m \mapsto \mathbb{R}$. Its j-th partial derivative at $x = (x_1, ..., x_m) \in \Omega$ is defined as

$$\frac{\partial f}{\partial x_j}(x) := \frac{\mathrm{d}f(x_1, \dots, x_{j-1}, t, x_{j+1}, \dots, x_m)}{\mathrm{d}t}|_{t=x_j}.$$
(17)

Exercise 10. Prove that $a_{ij} = \frac{\partial f_i}{\partial x_j}(x)$.

Note. This matrix (representation of the differential of f) is called the Jacobian matrix of f at x.

Exercise 11. Let $f: \mathbb{R}^m \mapsto \mathbb{R}^n$ be differentiable everywhere. Prove that if Df(x) = 0 for all $x \in \mathbb{R}^m$, then f is a constant.

• CHAIN RULE. Let $f(x): \mathbb{R}^m \mapsto \mathbb{R}^n$ and $g(y): \mathbb{R}^n \mapsto \mathbb{R}^k$. Then the composite function $g \circ f: \mathbb{R}^m \mapsto \mathbb{R}^k$ is defined as

$$(g \circ f)(x) = g(f(x)). \tag{18}$$

The partial derivatives of $g \circ f$ can be calculated through

$$\frac{\partial (g \circ f)_i}{\partial x_j} = \sum_{k=1}^n \frac{\partial g_i}{\partial y_k} (f(x)) \frac{\partial f_k(x)}{\partial x_j}.$$
(19)

Exercise 12. Find a calculus textbook and work on a few chain rule problems.

• HIGHER ORDER DERIVATIVES. We consider the simplest case $f: \Omega \subseteq \mathbb{R}^m \mapsto \mathbb{R}$. At every $x \in \Omega$, it has a differential which is a linear function $Df(x): \mathbb{R}^m \mapsto \mathbb{R}$. As this linear function can be represented by a $1 \times m$ matrix, it can also be identified with an *m*-vector. Thus at each point $x \in \Omega$, we have a vector $g(x) \in \mathbb{R}^m$. Differentiating this function g we obtain the second order derivative of f at x, denoted $D^2f(x)$.

Exercise 13. Prove or disprove: Let $x \in \Omega$ be arbitrary. The second order derivative of f can be seen as a bilinear form on $\mathbb{R}^{2m} = \mathbb{R}^m \times \mathbb{R}^m$, where each \mathbb{R}^m is the "velocity" space at x.

Exercise 14. Prove that the matrix representing the bilinear form is given by

$$a_{ij} = \frac{\partial^2 f}{\partial x_i \partial x_j} = \frac{\partial}{\partial x_i} \left(\frac{\partial f}{\partial x_j} \right). \tag{20}$$

As we can see here, for functions between \mathbb{R}^m and \mathbb{R}^n with m or n > 1, higher order derivatives get more and more complicated. However there is one exception.

Exercise 15. What happens if we take successive derivatives of a function $f: \mathbb{R} \mapsto \mathbb{R}^n$?

Note. The matrix representation of $D^2 f(x)$ is called the **Hessian matrix** of f at x.

Exercise 16. What are the Jacobian and Hessian matrices for $f: \mathbb{R} \mapsto \mathbb{R}$?

• TAYLOR EXPANSION.

Let $f: \Omega \subseteq \mathbb{R}^m \mapsto \mathbb{R}$. Then there holds

$$f(x+v) = f(x) + Df(x)(v) + \frac{1}{2}D^2f(x)(v,v) + R$$
(21)

where $\lim_{v\to 0} \frac{\|R\|}{\|v\|^2} = 0$. (21) is the Taylor expansion of f at x to second order.

Exercise 17. Prove the following.

$$Df(x)(v) = \sum_{i=1}^{m} \frac{\partial f(x)}{\partial x_i} v_i, \qquad D^2 f(x)(v,v) = \sum_{i=1}^{m} \sum_{j=1}^{m} \frac{\partial^2 f(x)}{\partial x_i \partial x_j} v_i v_j.$$
(22)

Examples.

Example 10. Let $f(x, y) := e^{xy}$. We calculate its Taylor expansion at (0, 0) to second order. We have

$$\begin{split} f(0,0) &= e^0 = 1;\\ \frac{\partial f}{\partial x} = y \, e^{xy} \Longrightarrow \frac{\partial f}{\partial x}(0,0) &= 0;\\ \frac{\partial f}{\partial y} = x \, e^{xy} \Longrightarrow \frac{\partial f}{\partial y}(0,0) &= 0;\\ \frac{\partial^2 f}{\partial x^2} = y^2 \, e^{xy} \Longrightarrow \frac{\partial^2 f}{\partial x^2}(0,0) &= 0;\\ \frac{\partial^2 f}{\partial x \partial y} = \frac{\partial^2 f}{\partial y \partial x} = e^{xy} \Longrightarrow \frac{\partial^2 f}{\partial x \partial y}(0,0) &= 1;\\ \frac{\partial^2 f}{\partial y^2} = x^2 \, e^{xy} \Longrightarrow \frac{\partial^2 f}{\partial y^2}(0,0) &= 0. \end{split}$$

Therefore the expansion reads

$$f(u,v) = f(0,0) + Df(0,0)(u,v) + \frac{1}{2}D^2f(0,0)(u,v) + R$$

= 1+uv + R. (23)

2. Linear algebra

The fundamental idea of calculus is to study functions through their first and second (or higher) derivatives. We have seen that such derivatives can be viewed as linear transformations between \mathbb{R}^m and \mathbb{R}^n . Therefore to understand them we need linear algebra.

2.1. Operations on matrices

- DETERMINANT.
 - Determinant is defined for square matrices.
 - \circ 2 × 2. Let $A := \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}$. Then

$$\det A := a_{11} a_{22} - a_{12} a_{21}. \tag{24}$$

Exercise 18. Let $u, v, x, y \in \mathbb{R}^3$. Prove that

$$(u \times v) \cdot (x \times y) = \det \left(\begin{array}{cc} u \cdot x & v \cdot x \\ u \cdot y & v \cdot y \end{array} \right).$$
(25)

$$\circ \quad 3 \times 3. \text{ Let } A := \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix}. \text{ Then}$$

det $A := a_{11} a_{22} a_{33} + a_{12} a_{23} a_{31} + a_{21} a_{32} a_{13} - a_{13} a_{22} a_{31} - a_{12} a_{21} a_{33} - a_{23} a_{32} a_{11}.$ (26)

Exercise 19. Let
$$u, v, w \in \mathbb{R}^3$$
. Let $A := \begin{pmatrix} u_1 & v_1 & w_1 \\ u_2 & v_2 & w_2 \\ u_3 & v_3 & w_3 \end{pmatrix}$. Prove that det $A = (u \times v) \cdot w$.

- A square matrix A is invertible, that is there is a matrix B such that $AB = BA = I = \begin{pmatrix} 1 & & \\ & & \\ & & 1 \end{pmatrix}$ if and only if det $A \neq 0$.
- EIGENVALUES AND EIGENVECTORS.
 - In most cases the matrix A under study is the mathematical representation of a linear function T.
 - We could try to understand an $n \times n$ matrix through its n^2 entries. However a more efficient way is through its eigenvalues/eigenvectors.
 - The eigenvalues of an $n \times n$ matrix A are the solutions to the equation

$$\det(\lambda I - A) = 0. \tag{27}$$

The eigenvectors corresponding to an eigenvalue λ_0 are those vectors satisfying

$$(\lambda I - A) v = 0 \Longleftrightarrow A v = \lambda v.$$
⁽²⁸⁾

• In many situations, one could choose n linearly independent eigenvectors $\{v_1, ..., v_n\}$ to form a "coordinate system", that is every vector $u \in \mathbb{R}^n$ can be uniquely represented as $u = a_1 v_1 + \cdots + a_n v_n$ where $a_1, ..., a_n \in \mathbb{R}$. Then the linear function T can be very easily understood:

$$T(u) = \lambda_1 a_1 v_1 + \dots + \lambda_n a_n v_n.$$
⁽²⁹⁾

- The most useful situation for us is when A is symmetric: $a_{ij} = a_{ji}$ for all i, j = 1, 2, ..., n. In this case we can choose n eigenvectors $v_1, ..., v_n$ satisfying
 - i. $v_i \perp v_j$, for all $i \neq j$;
 - ii. $||v_i|| = 1$ for all i = 1, 2, ..., n.

2.2. Linear dependence and independence

• LINEAR DEPENDENCE/INDEPENDENCE. Let $v_1, ..., v_k \in \mathbb{R}^n$. We say they are linearly dependent if there are $a_1, ..., a_k \in \mathbb{R}$, not all equal to 0, such that

$$a_1 v_1 + \dots + a_k v_k = 0. (30)$$

If no such a_1, \ldots, a_k exist, we say the vectors are linearly independent.

Exercise 20. Prove that, if $v_1, ..., v_k$ are linearly independent, then the following holds. If $a_1, ..., a_k \in \mathbb{R}$ satisfy $a_1 v_1 + \cdots + a_k v_k = 0$, then $a_1 = a_2 = \cdots = a_k = 0$.

Exercise 21. Prove that if $v_1, ..., v_k$ are linearly dependent, then at least one of them is "redundant" in the sense that it equals a linear combination of the other vectors.

• *n* vectors $v_1 = (v_{11}, ..., v_{1n}), ..., v_n = (v_{n1}, ..., v_{nn}) \in \mathbb{R}^n$ are linearly independent if and only if det $A \neq 0$ where

$$A = \left(\begin{array}{ccc} v_{11} & \cdots & v_{1n} \\ \vdots & \ddots & \vdots \\ v_{n1} & \cdots & v_{nn} \end{array}\right). \tag{31}$$

• n vectors $v_1, ..., v_n \in \mathbb{R}^n$ are linearly independent if and only if they form a "base" of \mathbb{R}^n , that is every $u \in \mathbb{R}^n$ can be written uniquely as a linear combination of $v_1, ..., v_n$.

3. Differential equations

A differential equation is an equation involving the derivative (and/or higher order derivatives) of functions.

3.1. The simplest and second simplest differential equations

In the following $x(t): \mathbb{R} \mapsto \mathbb{R}$ is a single variable function.

• The simplets ODE.

Consider

$$\frac{\mathrm{d}x}{\mathrm{d}t} = f(t). \tag{32}$$

Its solution, by definition, is the indefinite integral

$$x(t) = \int f(s) \,\mathrm{d}s \tag{33}$$

which by the fundamental theorem of calculus further equals

$$\int_0^t f(s) \,\mathrm{d}s + C \tag{34}$$

where C is an arbitrary constant.

If we further specify $x(t_0) = x_0$, then this constant is fixed.

Exercise 22. What is the solution to

$$\frac{\mathrm{d}x}{\mathrm{d}t} = f(t), \qquad x(t_0) = x_0? \tag{35}$$

Exercise 23. Solve

$$\frac{\mathrm{d}x}{\mathrm{d}t} = f(x), \qquad x(t_0) = x_0. \tag{36}$$

• THE NEXT SIMPLEST ODE.

Consider

$$\frac{\mathrm{d}x(t)}{\mathrm{d}t} + c x(t) = f(t), \qquad x(0) = x_0.$$
(37)

where $c \in \mathbb{R}$. Multiplying both sides by e^{ct} and apply Leibniz rule of differentiation of product of functions, we reach

$$\frac{\mathrm{d}}{\mathrm{d}t}(e^{ct}x(t)) = e^{ct}f(t),\tag{38}$$

which is the simplest ODE now. The solution of (38) is

$$e^{ct} x(t) = x_0 + \int_0^t e^{cs} f(s) \,\mathrm{d}s$$
 (39)

which now gives

$$x(t) = e^{-ct} x_0 + \int_0^t e^{c(s-t)} f(s) \,\mathrm{d}s.$$
(40)

Exercise 24. Solve

$$\frac{\mathrm{d}x(t)}{\mathrm{d}t} + c x(t) = f(t), \qquad x(t_0) = x_0.$$
(41)

In general, the equation

$$\frac{\mathrm{d}x(t)}{\mathrm{d}t} + c(t)x(t) = f(t) \tag{42}$$

could be solved by multiplying both sides by $e^{C(t)}$ where C(t) satisfies C'(t) = c(t).

Exercise 25. Solve

$$\frac{\mathrm{d}x(t)}{\mathrm{d}t} + c(t) x(t) = f(t), \qquad x(0) = x_0.$$
(43)

3.2. Existence and uniqueness of a general first order ODE

- General situation we face in 348.
 - In 348 we often need to study systems of differential equations:

$$\frac{\mathrm{d}x_1(t)}{\mathrm{d}t} = f_1(x_1(t), \dots, x_n(t)), \qquad x_1(0) = x_{01}$$

$$\vdots \quad \vdots \quad \vdots$$
(44)

$$\frac{\mathrm{d}x_n(t)}{\mathrm{d}t} = f_n(x_1(t), \dots, x_n(t)). \qquad x_n(0) = x_{0n}$$
(45)

• Taking advantage of multivariable calculus, we can re-write the above to a more compact form:

$$\frac{\mathrm{d}x(t)}{\mathrm{d}t} = f(x(t)), \qquad x(0) = x_0 \tag{46}$$

where $x: \mathbb{R} \mapsto \mathbb{R}^n$, $f: \mathbb{R}^n \mapsto \mathbb{R}^n$, and $x_0 = (x_{01}, \dots, x_{0n})$.

- The one ODE system that we could really solve.
 - For general f there is little hope explicitly solving (46). However there is a special, very useful, situation that we could solve.
 - Consider the case where

$$f_i(x_1, \dots, x_n) = a_{i1}x_1 + \dots + a_{in}x_n, \qquad i = 1, 2, \dots, n$$
(47)

where $a_{ij} \in \mathbb{R}$ for all i, j = 1, 2, ..., n. In other words, we have

$$\frac{\mathrm{d}x(t)}{\mathrm{d}t} = A x(t), \qquad x(0) = x_0.$$
(48)

• Now to further simplify the situation, we assume A has n linearly independent eigenvectors v_1, \ldots, v_n . Writing $x(t) = z_1(t) v_1 + \cdots + z_n(t) v_n$ and $x_0 = z_{01}v_1 + \cdots + z_{0n}v_n$ we obtain

$$\frac{\mathrm{d}z_i(t)}{\mathrm{d}t} = \lambda_i \, z_i(t), \qquad z_i(0) = z_{i0}. \tag{49}$$

Exercise 26. Justify (49).

Exercise 27. Finish the solution.

• EXISTENCE AND UNIQUENESS THEOREM.

Although there is little hope explicitly solving (46), we have the following qualitative understanding.

THEOREM 11. Let $f: \mathbb{R}^n \mapsto \mathbb{R}^n$ be differentiable with continuous derivatives. Then there is T > 0 and a unique function x(t) satisfying $x(0) = x_0$, and for all 0 < t < T,

$$\frac{\mathrm{d}x(t)}{\mathrm{d}t} = f(x). \tag{50}$$

Exercise 28. Let $A(t) = \begin{pmatrix} a_{11}(t) & \cdots & a_{1n}(t) \\ \vdots & \ddots & \vdots \\ a_{n1}(t) & \cdots & a_{nn}(t) \end{pmatrix}$ be such that $a_{ij}(t)$ are continuous functions of t for every i, j = 1, 2, ..., n. Prove that the solution to

$$\frac{\mathrm{d}x(t)}{\mathrm{d}t} = A(t) x(t) \tag{51}$$

exists and is unique.

Remark 12. We will see later that existence of curves with certain desirable properties would be equivalent to the existence of solution to an ODE of the form (46). Then Theorem 11 would guarantee us that this curve exists.