Math 2250-004

Week 8: Finish sections 4.2-4.4 and linear combination concepts, and then begin Chapter 5 on linear differential equations, sections 5.1-5.2.

Mon Feb 27: Use last Friday's notes to talk about linear independence/dependence, span, subspaces and bases. That discussion will continue into these notes. Here is a summary of vocabulary terms so far:

A <u>linear combination</u> of the vectors $\underline{v}_1, \underline{v}_2, \dots \underline{v}_n$ is

any vector \underline{v} that is a sum of scalar multiples of those vectors, i.e. any \underline{v} expressible as $\underline{\mathbf{y}} = c_1 \underline{\mathbf{y}}_1 + c_2 \underline{\mathbf{y}}_2 + \dots + c_n \underline{\mathbf{y}}_n$

The <u>span</u> of $\{\underline{v}_1, \underline{v}_2, \dots \underline{v}_n\}$ is

the collection of all possible linear combinations:
$$span\{\underline{\mathbf{v}}_1,\underline{\mathbf{v}}_2, \dots \underline{\mathbf{v}}_n\} := \{\underline{\mathbf{v}} = c_1\underline{\mathbf{v}}_1 + c_2\underline{\mathbf{v}}_2 + \dots + c_n\underline{\mathbf{v}}_n \text{ such that each } c_i \in \mathbb{R}, i = 1, 2, \dots n\}$$

The vectors $\underline{v}_1, \underline{v}_2, \dots \underline{v}_n$ are <u>linearly independent</u> means

$$c_1 \underline{\mathbf{y}}_1 + c_2 \underline{\mathbf{y}}_2 + \dots + c_n \underline{\mathbf{y}}_n = 0 \Rightarrow c_1 = c_2 = \dots = c_n = 0.$$

(Equivalently, no \underline{v}_i can be expressed as a linear combination of some of the other vectors in the collection.)

The vectors $\underline{v}_1, \underline{v}_2, \dots \underline{v}_n$ are <u>linearly dependent</u> means

they are not linearly independent.

(Equivalently, some \underline{v}_i can be expressed as a linear combination of some of the other vectors in the collection.)

In Friday's notes but not yet discussed before Monday:

A subset W of \mathbb{R}^n is a <u>subspace</u> means

that W is closed under addition and scalar multiplication:

$$(\alpha) \qquad \underline{u}, \underline{v} \in W \Rightarrow \underline{u} + \underline{v} \in W,$$

$$(\beta) \qquad \underline{u} \in W, c \in \mathbb{R} \Rightarrow c \,\underline{u} \in W.$$

Let W be a subspace. A basis for W is

a collection of vectors $\underline{\mathbf{v}}_1, \underline{\mathbf{v}}_2, \dots \underline{\mathbf{v}}_n$ that are linearly independent and span W.

Equivalently, each $\underline{\mathbf{w}} \in W$ can be written as $\underline{\mathbf{w}} = c_1 \underline{\mathbf{v}}_1 + c_2 \underline{\mathbf{v}}_2 + \dots + c_n \underline{\mathbf{v}}_n$ for exactly one choice of linear combination coefficients $c_1, c_2, ...c_n$.

The <u>dimension</u> of a subspace W is the number of vectors in a basis for W. (It turns out that all bases for a subspace always have the same number of vectors.)

Note: Subspaces are special subsets because they are closed with respect to all linear combinations (since linear combinations are built up by successive scalar multiplication and addition operations). The reason why we use the word "subspaces" for these special subsets is because there is a more general notion of vector space for collections of objects that can be added and scalar multiplied so that the usual addition and scalar multiplication axioms hold. \mathbb{R} , \mathbb{R}^2 , \mathbb{R}^3 , .. \mathbb{R}^m are vector spaces. Subspaces of \mathbb{R}^m are also vector spaces in their own right. In Chapter 5 we'll be focusing on function vector spaces (since you can add and scalar multiply functions), and subspaces of those. The purpose will be to understand the solution spaces for higher order linear differential equations, and applications.

Exercise 1) Most <u>subsets</u> of \mathbb{R}^m are actually not <u>subspaces</u>. Show that

- <u>1a)</u> $W = \{ [x, y]^T \in \mathbb{R}^2 \text{ s.t. } x^2 + y^2 = 4 \} \text{ is } \underline{\text{not a subspace of } } \mathbb{R}^2 .$

- 1b) $W = \{ [x, y]^T \in \mathbb{R}^2 \text{ s.t. } y = 3 \text{ x} + 1 \}$ is <u>not</u> a subspace of \mathbb{R}^2 . 1c) $W = \{ [x, y]^T \in \mathbb{R}^2 \text{ s.t. } y = 3 \text{ x} \}$ is a subspace of \mathbb{R}^2 . Then find a basis for this subspace. 1d) $W = \{ [x, y]^T \in \mathbb{R}^2 \text{ s.t. } y = 3 \text{ x} \}$ is a subspace of \mathbb{R}^2 . Then find a basis for this subspace. 1d) $W = \{ [x, y]^T \in \mathbb{R}^2 \text{ s.t. } y = 3 \text{ x} \}$ is a non-homogeneous matrix equation $A \underline{x} = \underline{b} \ (\underline{b} \neq \underline{0})$, with A_{mxn} is not a subspace. In particular, lines and planes that don't go through the origin are not subspaces.

Key facts about how subspaces DO arise:

There are two ways that subspaces arise: (These ideas will be important when we return to differential equations, in Chapter 5, although it's probably difficult to envision what they have to do with differential equations right now.)

1)
$$W = span\{\underline{\mathbf{v}}_1, \underline{\mathbf{v}}_2, \dots, \underline{\mathbf{v}}_n\}.$$

Expressing a subspace this way is an <u>explicit</u> way to describe the subspace W, because you are "listing" all of the vectors in it. In this case we prefer that $\underline{v}_1, \underline{v}_2, \dots, \underline{v}_n$ be linearly independent, i.e. a basis, because that guarantees that each $\underline{w} \in W$ is a unique linear combination of these spanning vectors.

Recall why W is a subspace: Let \underline{v} , $\underline{w} \in W \Rightarrow$

$$\underline{\mathbf{y}} = c_1 \underline{\mathbf{y}}_1 + c_2 \underline{\mathbf{y}}_2 + \dots + c_n \underline{\mathbf{y}}_n$$

$$\underline{\mathbf{w}} = d_1 \underline{\mathbf{y}}_1 + d_2 \underline{\mathbf{y}}_2 + \dots + d_n \underline{\mathbf{y}}_n$$

$$\Rightarrow \underline{\mathbf{y}} + \underline{\mathbf{w}} = \left(c_1 + d_1\right) \underline{\mathbf{y}}_1 + \left(c_2 + d_2\right) \underline{\mathbf{y}}_2 + \dots + \left(c_n + d_n\right) \underline{\mathbf{y}}_n \in W \text{ (verifies } \alpha)$$
and let $c \in \mathbb{R} \Rightarrow$

$$c\underline{\mathbf{y}} = cc_1 \underline{\mathbf{y}}_1 + cc_2 \underline{\mathbf{y}}_2 + \dots + cc_n \underline{\mathbf{y}}_n \in W \text{ (verifies } \beta)$$

2)
$$W = \{ \underline{x} \in \mathbb{R}^n \text{ such that } A_{m \times n} \underline{x} = \underline{\mathbf{0}} \}.$$

This is an <u>implicit way</u> to describe the subspace W because you're only specifying a homogeneous matrix equation that the vectors in W must satisfy, but you're not saying what the vectors are.

Why W is a subspace: Let $\underline{\mathbf{v}}$, $\underline{\mathbf{w}} \in W \Rightarrow$

$$A\underline{\mathbf{v}} = \underline{\mathbf{0}}, A\underline{\mathbf{w}} = \underline{\mathbf{0}} \Rightarrow A\underline{\mathbf{v}} + A\underline{\mathbf{w}} = \underline{\mathbf{0}} \Rightarrow A(\underline{\mathbf{v}} + \underline{\mathbf{w}}) = \underline{\mathbf{0}} \Rightarrow \underline{\mathbf{v}} + \underline{\mathbf{w}} \in W \text{ (verifies } \alpha)$$

and let $c \in \mathbb{R} \Rightarrow$

$$A\underline{\mathbf{v}} = \underline{\mathbf{0}} \Rightarrow cA\underline{\mathbf{v}} = c\underline{\mathbf{0}} = \underline{\mathbf{0}} \Rightarrow A(c\underline{\mathbf{v}}) = \underline{\mathbf{0}} \Rightarrow c\underline{\mathbf{v}} \in W \text{ (verfies } \beta).$$

Example: Last week we saw that

(1)
$$W = span\{\underline{v}_1, \underline{v}_2\}$$
 with

$$W = span \left\{ \underline{\boldsymbol{v}}_{1} = \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix}, \underline{\boldsymbol{v}}_{2} = \begin{bmatrix} -1 \\ 2 \\ 0 \end{bmatrix} \right\}$$

is the collection of position vectors for points (x, y, z) satisfying the implicit equation (2)

$$-2x - y + z = 0$$
.

In other words the explicit description of W as the span of two vectors in (1) corresponds to the implicit description of W as the space of solutions to the homogeneous matrix equation

$$\begin{bmatrix} -2 & -1 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 0 \end{bmatrix}.$$

Tues Feb 28

Continuing our discussion of linear combinations, span, linear independence/dependence, subspace, basis, dimension.

Exercise 1) Use geometric reasoning to argue why the <u>only</u> subspaces of \mathbb{R}^2 are

- $\overline{(0)}$ The single vector $[0, 0]^T$, or
- (1) A line through the origin, i.e. $span\{\underline{u}\}$ for some non-zero vector \underline{u} , or
- (2) All of \mathbb{R}^2 .

Exercise 2) Use matrix theory to show that the only subspaces of \mathbb{R}^3 are

- (0) The single vector $[0, 0, 0]^T$, or
- (1) A line through the origin, i.e. $span\{\underline{u}\}$ for some non-zero vector \underline{u} , or
- (2) A plane through the origin, i.e. $span\{\underline{u},\underline{v}\}$ where $\underline{u},\underline{v}$ are linearly independent, or
- (3) All of \mathbb{R}^3 .

Exercise 3) What are the dimensions of the subspaces in Exercise 4 and Exercise 5? How do these ideas generalize to subspaces of \mathbb{R}^n ?

Usually in applications we do not start with a basis for a subspace - rather this is the goal we search for, since the entire subspace may be reconstructed explicitly and precisely from the basis (which is why a basis is called "a basis"). Usually, our subspace W in \mathbb{R}^m is likely to be described in an implicit manner, as the solution space to a homogeneous matrix equation. In Chapter 5 the subspaces W will be the solution spaces to "homogeneous linear differential equations."

If we wish to find a basis for the homogeneous solution space $W = \{\underline{x} \in \mathbb{R}^n \text{ such that } A_{m \times n} \underline{x} = \underline{0} \}$, then the following algorithm will always work: reduce the augmented matrix, backsolve and write the explicit solution in linear combination form. The vectors that you are taking linear combinations of will always span the solution space, by construction. If you follow this algorithm they will automatically be linearly independent, so they will be a basis for the solution space. This is illustrated in the large example below:

Exercise 4 Consider the matrix equation $A \underline{x} = \underline{0}$, with the matrix A (and its reduced row echelon form) shown below:

Find a basis for the solution space $W = \{\underline{x} \in \mathbb{R}^6 \text{ s.t. } A \underline{x} = \underline{\mathbf{0}}\}$ by backsolving, writing your explicit solutions in linear combination form, and extracting a basis. Explain why these vectors span the solution space and verify that they're linearly independent.

Solution (don't peek:-): backsolve, realizing that the augmented matrices have final columns of zero. Label the free variables with the letter "t", and subscripts to show the non-leading 1 columns from which they arose:

$$x_6 = t_6, x_5 = t_5, x_4 = t_4, x_3 = -2 t_4 + t_5 - 3 t_6$$

 $x_2 = t_2, x_1 = -2 t_2 - t_4 - t_5 - 2 t_6$.

In vector form and then linear combination form this is:

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \end{bmatrix} = \begin{bmatrix} -2 t_2 - t_4 - t_5 - 2 t_6 \\ & t_2 \\ & -2 t_4 + t_5 - 3 t_6 \\ & t_4 \\ & t_5 \\ & t_6 \end{bmatrix} = t_2 \begin{bmatrix} -2 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} + t_4 \begin{bmatrix} -1 \\ 0 \\ -2 \\ 1 \\ 0 \\ 0 \end{bmatrix} + t_5 \begin{bmatrix} -1 \\ 0 \\ 1 \\ 0 \\ 0 \end{bmatrix} + t_6 \begin{bmatrix} -2 \\ 0 \\ -3 \\ 0 \\ 1 \end{bmatrix}.$$

Thus the four vectors

$$\begin{bmatrix} -2 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ -2 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 1 \\ 0 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} -2 \\ 0 \\ -3 \\ 0 \\ 0 \\ 1 \end{bmatrix}$$

are a basis for the solution space:

- They span the solution space by construction.
- They are linearly independent because if we set a linear combination equal to zero:

$$\begin{bmatrix} -2 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} + t_4 \begin{bmatrix} -1 \\ 0 \\ -2 \\ 1 \\ 0 \\ 0 \end{bmatrix} + t_5 \begin{bmatrix} -1 \\ 0 \\ 1 \\ 0 \\ 1 \\ 0 \end{bmatrix} + t_6 \begin{bmatrix} -2 \\ 0 \\ -3 \\ 0 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} -2 t_2 - t_4 - t_5 - 2 t_6 \\ t_2 - t_4 - t_5 - 2 t_6 \\ -2 t_4 + t_5 - 3 t_6 \\ t_4 - t_5 - 3 t_6 \\ t_5 - t_6 \end{bmatrix}$$

then looking in the second entry implies $t_2 = 0$, the fourth entry implies $t_4 = 0$, and similarly $t_5 = t_6 = 0$.

Exercise 5) Focus on the idea that solutions to homogeneous matrix equations correspond exactly to linear dependencies between the columns of the matrix. (If it helps, think of renaming the vector \underline{x} in the example above, with a vector \underline{c} of linear combination coefficients; then recall the prime Chapter 4 algebra fact that

$$A \underline{c} = c_1 col_1(A) + c_2 col_2(A) + \dots + c_n col_n(A)$$

so any solution \underline{c} to $A\underline{c} = \underline{\theta}$ is secretly a columns dependency, and vise-verse.)

Now, since the solution set to a homogeneous linear system does not change as you do elementary row operations to the augmented matrix, column dependencies also do not change. Therefore your basis in Exercise 4 for the homogeneous solution space in \mathbb{R}^6 can also be thought of as a "basis" of the key column dependencies in \mathbb{R}^4 , for both the original matrix, and for the reduced row echelon form. 5a) Check this, by reading off "easy" column dependencies in the reduced matrix; seeing that they are also dependencies in the original matrix; and that they correspond to the basis of the homogeneous solution space. Magic! We will use this magic in important interesting ways, later in the course.

<u>5b)</u> Find a basis for W = the span of the five columns of A. Hint: this will be a two-dimensional subspace of \mathbb{R}^4 and cyou can create it by successively removing redundant (dependent) vectors from the original collection of the five column vectors, until your remaining set still spans W but is linearly independent.

Exercise 6) (This exercise explains why any given matrix has only one reduced row echelon form, no matter what sequence of elementary row operations one uses to find it. We didn't have the tools to explain why this fact was true earlier, back in Chapter 3.) Let $B_{4 \times 5}$ be a matrix whose columns satisfy the following dependencies:

$$col_1(B) \neq \underline{\mathbf{0}} \quad \text{(i.e. is independent)}$$

$$col_2(B) = 3 \ col_1(B)$$

$$col_3(B) \text{ is independent of column 1}$$

$$col_4(B) \text{ is independent of columns 1,3.}$$

$$col_5(B) = -3 \ col_1(B) + 2 \ col_3(B) - col_4(B).$$

What is the reduced row echelon form of *B*?

Some important facts about spanning sets, independence, bases and dimension follow from one key fact, and then logic. We will want to use these facts going forward, as we return to studying differential equations tomorrow.

<u>key fact</u>: If *n* vectors \underline{v}_1 , \underline{v}_2 , ... \underline{v}_n span a subspace *W* then any collection \underline{w}_1 , \underline{w}_2 , ... \underline{w}_N of vectors in *W* with N > n will always be linearly dependent. (This is explained on pages 254-255 of the text, and has to to do with matrix facts that we already know.) Notice too that this fact fits our intuition based on what we know in the special cases that we've studied, in particular $W = \mathbb{R}^n$.)

Thus:

- 1) If a finite collection of vectors in W is linearly independent, then no collection with fewer vectors can span all of W. (This is because if the smaller collection did span, the larger collection wouldn't have been linearly independent after all, by the <u>key fact.</u>)
- <u>2)</u> Every basis of W has the same number of vectors, so the concept of dimension is well-defined and doesn't depend on choice of basis. (This is because if $\underline{v}_1, \underline{v}_2, \dots \underline{v}_n$ are a basis for W then every larger collection of vectors is dependent by the <u>key fact</u> and every smaller collection fails to span by (1), so only collections with exactly n vectors have a chance to be bases.)
- 3) Let the dimension of W be the number n, i.e. there is some basis $\underline{v}_1, \underline{v}_2, \dots \underline{v}_n$ for W. Then if vectors $\underline{w}_1, \underline{w}_2, \dots \underline{w}_n$ span W then they're automatically linearly independent and thus a basis. (If they were dependent we could delete one of the \underline{w}_j that was a linear combination of the others and still have a spanning set. This would violate (1) since $\underline{v}_1, \underline{v}_2, \dots \underline{v}_n$ are linearly independent.)
- 4) Let the dimension of W be the number n, i.e. there is some basis $\underline{v}_1, \underline{v}_2, \dots \underline{v}_n$ for W. Then if $\underline{w}_1, \underline{w}_2, \dots \underline{w}_n$ are in W and are linearly independent, they automatically span W and thus are a basis. (If they didn't span W we could augment with a vector \underline{w}_{n+1} not in their span and have a collection of n+1 still independent* vectors in W, violating the <u>key fact.</u>
- * Check: If $\underline{w}_1, \underline{w}_2, \dots \underline{w}_n$ are linearly independent, and $\underline{w}_{n+1} \notin span\{\underline{w}_1, \underline{w}_2, \dots \underline{w}_n\}$, then $\underline{w}_1, \underline{w}_2, \dots \underline{w}_n, \underline{w}_{n+1}$ are also linearly independent. This fact generalizes the ideas we used when we figured out all possible subspaces of \mathbb{R}^3 . Here's how it goes:

To show the larger collection is still linearly independent study the equation

$$c_1 \underline{\mathbf{w}}_1 + c_2 \underline{\mathbf{w}}_2 + \dots + c_n \underline{\mathbf{w}}_n + d \underline{\mathbf{w}}_{n+1} = \underline{\mathbf{0}}.$$

Since $\underline{w} \notin span\{\underline{w}_1, \underline{w}_2, \dots \underline{w}_n\}$ it must be that d = 0 (since otherwise we could solve for \underline{w}_{n+1} as a linear combination of $\underline{w}_1, \underline{w}_2, \dots \underline{w}_n$). But once d = 0, we have

$$c_1 \underline{\mathbf{w}}_1 + c_2 \underline{\mathbf{w}}_2 + \dots + c_n \underline{\mathbf{w}}_n = \underline{\mathbf{0}}$$

which implies $c_1 = c_2 = ... = c_n = 0$ by the independence of $\underline{\boldsymbol{w}}_1, \underline{\boldsymbol{w}}_2, ... \underline{\boldsymbol{w}}_n$.

Math 2250-004 Wed Mar 1

5.1 Second order linear differential equations, and vector space theory connections.

Definition: A <u>vector space</u> is a collection of objects together with and "addition" operation "+", and a scalar multiplication operation, so that the rules below all hold.

- (a) Whenever $f, g \in V$ then $f + g \in V$. (closure with respect to addition)
- (β) Whenever $f \in V$ and $c \in \mathbb{R}$, then $c \cdot f \in V$. (closure with respect to scalar

multiplication)

As well as:

- (a) f + g = g + f (commutative property)
- (b) f + (g + h) = (f + g) + h (associative property)
- (c) $\exists 0 \in V$ so that f + 0 = f is always true.
- (d) $\forall f \in V \exists -f \in V \text{ so that } f + (-f) = 0 \text{ (additive inverses)}$
- (e) $c \cdot (f+g) = c \cdot f + c \cdot g$ (scalar multiplication distributes over vector addition)
- (f) $(c_1 + c_2) \cdot f = c_1 \cdot f + c_2 \cdot f$ (scalar addition distributes over scalar multiplication)
- (g) $c_1 \cdot (c_2 \cdot f) = (c_1 c_2) \cdot f$ (associative property)
- (h) $1 \cdot f = f$, $(-1) \cdot f = -f$, $0 \cdot f = 0$ (these last two actually follow from the others).

Examples we've seen:

- (1) \mathbb{R}^m , with the usual vector addition and scalar multiplication, defined component-wise
- (2) subspaces W of \mathbb{R}^m , which satisfy (α) , (β) , and therefore automatically satisfy (a)-(h), because the vectors in W also lie in \mathbb{R}^m .

Exercise 0) In Chapter 5 we focus on the vector space

$$V = C(\mathbb{R}) := \{ f : \mathbb{R} \to \mathbb{R} \text{ s.t. } f \text{ is a continuous function} \}$$

and its subspaces. Verify that the vector space axioms for linear combinations are satisfied for this space of functions. Recall that the function f + g is defined by (f + g)(x) := f(x) + g(x) and the scalar multiple cf(x) is defined by (cf)(x) := cf(x). What is the zero vector for functions?

Because the vector space axioms are exactly the arithmetic rules we used to work with linear combination equations, all of the concepts and vector space theorems we talked about for \mathbb{R}^m and its subspaces make sense for the function vector space V and its subspaces. In particular we can talk about

- the span of a finite collection of functions $f_1, f_2, ... f_n$.
- linear independence/dependence for a collection of functions $f_1, f_2, \dots f_n$.
- subspaces of V
- bases and dimension for finite dimensional subspaces. (The function space V itself is infinite dimensional, meaning that no finite collection of functions spans it.)

<u>Definition:</u> A <u>second order linear</u> differential equation for a function y(x) is a differential equation that can be written in the form

$$A(x)y'' + B(x)y' + C(x)y = F(x)$$
.

We search for solution functions y(x) defined on some specified interval I of the form a < x < b, or (a, ∞) , $(-\infty, a)$ or (usually) the entire real line $(-\infty, \infty)$. In this chapter we assume the function $A(x) \neq 0$ on I, and divide by it in order to rewrite the differential equation in the standard form

$$y'' + p(x)y' + q(x)y = f(x)$$
.

One reason this DE is called <u>linear</u> is that the "operator" L defined by

$$L(y) := y'' + p(x)y' + q(x)y$$

satisfies the so-called <u>linearity properties</u>

$$(1) L(y_1 + y_2) = L(y_1) + L(y_2)$$

$$(2) L(cy) = cL(y), c \in \mathbb{R}.$$

(Recall that the matrix multiplication function $L(\underline{x}) := A \underline{x}$ satisfies the analogous properties. Any time we have have a transformation L satisfying (1),(2), we say it is a <u>linear transformation</u>.)

Exercise 1a) Check the linearity properties (1), (2) for the differential operator L.

1b) Use these properties to show that

Theorem 0: the solution space to the <u>homogeneous</u> second order linear DE

$$y'' + p(x)y' + q(x)y = 0$$

is a subspace. Notice that this is the "same" proof we used Monday to show that the solution space to a homogeneous matrix equation is a subspace.

Exercise 2) Find the solution space to homogeneous differential equation for y(x)

$$v'' + 2v' = 0$$

on the *x*-interval $-\infty < x < \infty$. Notice that the solution space is the span of two functions. Hint: This is really a first order DE for v = y'.

Exercise 3) Use the linearity properties to show

Theorem 1: All solutions to the <u>nonhomogeneous</u> second order linear DE

$$y'' + p(x)y' + q(x)y = f(x)$$

are of the form $y = y_P + y_H$ where y_P is any single particular solution and y_H is some solution to the homogeneous DE. (y_H is called y_c , for complementary solution, in the text). Thus, if you can find a single particular solution to the nonhomogeneous DE, and all solutions to the homogeneous DE, you've actually found all solutions to the nonhomogeneous DE. (You had a homework problem related to this idea, 3.4.40, in homework 5, but in the context of matrix equations, a week or two ago. The same idea reappears in this week's lab 6, in problem 1d!)

Exercise 4) Verify Theorem 1 for the differential equation

$$y'' + 2y' = 3$$

Theorem 2 (Existence-Uniqueness Theorem): Let p(x), q(x), f(x) be specified continuous functions on the interval I, and let $x_0 \in I$. Then there is a unique solution y(x) to the <u>initial value problem</u>

$$y'' + p(x)y' + q(x)y = f(x)$$

$$y(x_0) = b_0$$

$$y'(x_0) = b_1$$

and y(x) exists and is twice continuously differentiable on the entire interval I.

Exercise 5) Verify Theorem 2 for the interval $I = (-\infty, \infty)$ and the IVP

$$y'' + 2y' = 3$$

 $y(0) = b_0$
 $y'(0) = b_1$

Unlike in the previous example, and unlike what was true for the first order linear differential equation

$$y' + p(x)y = q(x)$$

there is <u>not</u> a clever integrating factor formula that will always work to find the general solution of the second order linear differential equation

$$y'' + p(x)y' + q(x)y = f(x)$$
.

Rather, we will usually resort to vector space theory and algorithms based on clever guessing. It will help to know

Theorem 3: The solution space to the second order homogeneous linear differential equation

$$y'' + p(x)y' + q(x)y = 0$$

is 2-dimensional.

This Theorem is illustrated in <u>Exercise 2</u> that we completed earlier. The theorem <u>and</u> the techniques we'll actually be using going forward are illustrated by

Exercise 5) Consider the homogeneous linear DE for y(x)

$$y'' - 2y' - 3y = 0$$

<u>5a</u>) Find two exponential functions $y_1(x) = e^{rx}$, $y_2(x) = e^{\rho x}$ that solve this DE.

5b) Show that every IVP

$$y'' - 2y' - 3y = 0$$

 $y(0) = b_0$
 $y'(0) = b_1$

can be solved with a unique linear combination $y(x) = c_1 y_1(x) + c_2 y_2(x)$.

Then use the uniqueness theorem to deduce that y_1, y_2 span the solution space to this homogeneous differential equation.

Next, show $y_1(x) = e^{rx}$, $y_2(x) = e^{\rho x}$ are linearly independent by setting a linear combination equal to the zero function, differentiating that identity, and then substituting x = 0 into the resulting system of equations to deduce $c_1 = c_2 = 0$:

$$c_1 y_1(x) + c_2 y_2(x) = 0$$

 $\Rightarrow c_1 y_1'(x) + c_2 y_2'(x) = 0$

so that $\{y_1(x) = e^{rx}, y_2(x) = e^{\rho x}\}$ is a basis for the solution space. So also the solution space is two-dimensional since the basis consists of two functions.

<u>5c</u>) Now consider the inhomogeneous DE

$$y'' - 2y' - 3y = 9$$

Notice that $y_P(x) = -3$ is a particular solution. Use this information and superposition (linearity) to solve the initial value problem

$$y'' - 2y' - 3y = 9$$

 $y(0) = 6$
 $y'(0) = -2$.

First, finish and/or review our Wednesday discussions.

Although we don't have the tools yet to prove the existence-uniqueness result Theorem 2, we can use it to prove the dimension result Theorem 3. Here's how (and this is really just an abstractified version of Exercise 5 in Wednesday's notes).

Consider the homogeneous differential equation

$$y'' + p(x)y' + q(x)y = 0$$

on an interval I for which the hypotheses of the existence-uniqueness theorem hold. Pick any $x_0 \in I$. Find solutions $y_1(x)$, $y_2(x)$ to IVP's at x_0 so that the so-called Wronskian matrix for $y_1, y_2 \text{ at } x_0$

$$W(y_1, y_2)(x_0) = \begin{bmatrix} y_1(x_0) & y_2(x_0) \\ y_1'(x_0) & y_2'(x_0) \end{bmatrix}$$

 $W\left(y_1,y_2\right)\left(x_0\right) = \begin{bmatrix} y_1\left(x_0\right) & y_2\left(x_0\right) \\ y_1{'}\left(x_0\right) & y_2{'}\left(x_0\right) \end{bmatrix}$ is invertible (i.e. $\begin{bmatrix} y_1\left(x_0\right) \\ y_1{'}\left(x_0\right) \\ y_2{'}\left(x_0\right) \end{bmatrix}$ are a basis for \mathbb{R}^2 , or equivalently so that the determinant of

the Wronskian matrix (called just the Wronskian) is non-zero at x_0).

You may be able to find suitable y_1, y_2 by good guessing, as in Exercise 5 on Friday, but the existence-uniqueness theorem guarantees they exist even if you don't know how to find formulas for them.

Under these conditions, the solutions y_1, y_2 are actually a <u>basis</u> for the solution space! Here's why: • span: the condition that the Wronskian matrix is invertible at x_0 means we can solve each IVP there with a linear combination $y = c_1 y_1 + c_2 y_2$: In that case, $y' = c_1 y_1' + c_2 y_2'$ so to solve the IVP

$$y'' + p(x)y' + q(x)y = 0$$

 $y(x_0) = b_0$
 $y'(x_0) = b_1$

we set

$$y(x) = c_1 y_1(x) + c_2 y_2(x).$$

At x_0 we wish to find c_1 , c_2 so that

$$c_1 y_1(x_0) + c_2 y_2(x_0) = b_0$$

 $c_1 y_1'(x_0) + c_2 y_2'(x_0) = b_1$

Since the Wronskian matrix at x_0 has and inverse, the unique solution $\begin{bmatrix} c_1, c_2 \end{bmatrix}^T$ is given by

$$\begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} y_1(x_0) & y_2(x_0) \\ y_1'(x_0) & y_2'(x_0) \end{bmatrix}^{-1} \begin{bmatrix} b_0 \\ b_1 \end{bmatrix}.$$

Since the uniqueness theorem says each IVP has a unique solution, this must be it. Since each solution

y(x) to the differential equation solves *some* initial value problem at x_0 , each solution y(x) is a linear combination of y_1, y_2 . Thus y_1, y_2 span the solution space.

• Linear independence: If we have the identity

$$c_1 y_1(x) + c_2 y_2(x) = 0$$

then by differentiating each side with respect to x we also have

$$c_1 y_1'(x) + c_2 y_2'(x) = 0.$$

Evaluating at $x = x_0$ this is the system

$$c_1 y_1(x_0) + c_2 y_2(x_0) = 0$$

$$c_1 y_1'(x_0) + c_2 y_2'(x_0) = 0$$

so

$$\begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} y_1(x_0) & y_2(x_0) \\ y_1'(x_0) & y_2'(x_0) \end{bmatrix}^{-1} \begin{bmatrix} 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

5.2: general theory for n^{th} -order linear differential equations; tests for linear independence; also begin 5.3: finding the solution space to homogeneous linear constant coefficient differential equations by trying exponential functions as potential basis functions.

The two main goals in Chapter 5 are to learn the structure of solution sets to n^{th} order linear DE's, including how to solve the IVPs

$$y^{(n)} + a_{n-1}y^{(n-1)} + \dots + a_1y' + a_0y = f$$

$$y(x_0) = b_0$$

$$y'(x_0) = b_1$$

$$y''(x_0) = b_2$$

$$\vdots$$

$$y^{(n-1)}(x_0) = b_{n-1}$$

and to learn important physics/engineering applications of these general techniques.

The algorithm for solving these DEs and IVPs is:

- (1) Find a basis $y_1, y_2, ..., y_n$ for the *n*-dimensional homogeneous solution space, so that the general homogeneous solution is their span, i.e. $y_H = c_1 y_1 + c_2 y_2 + ... + c_n y_n$.
- (2) If the DE is non-homogeneous, find a particular solution y_P . Then the general solution to the non-homogeneous DE is $y = y_P + y_H$. (If the DE is homogeneous you can think of taking $y_P = 0$, since $y = y_H$.)
- (3) Find values for the *n* free parameters $c_1, c_2,...c_n$ in

$$y = y_P + c_1 y_1 + c_2 y_2 + ... + c_n y_n$$

to solve the initial value problem with initial values b_0 , b_1 ,..., b_{n-1} . (This last step just reduces to a matrix problem like in Chapter 3, where the matrix is the Wronskian matrix of y_1 , y_2 ,... y_n , evaluated at x_0 and the right hand side vector comes from the initial values and the particular solution and its derivatives' values at x_0 .)

We've already been exploring how these steps play out in examples and homework problems, but will be studying them more systematically today and Monday. On Tuesday we'll begin the applications in section 5.4. We should have some fun experiments later next week to compare our mathematical modeling with physical reality.

<u>Definition</u>: An n^{th} order linear differential equation for a function y(x) is a differential equation that can be written in the form

$$A_n(x)y^{(n)} + A_{n-1}(x)y^{(n-1)} + \dots + A_1(x)y' + A_0(x)y = F(x)$$
.

We search for solution functions y(x) defined on some specified interval I of the form a < x < b, or (a, ∞) , $(-\infty, a)$ or (usually) the entire real line $(-\infty, \infty)$. In this chapter we assume the function $A_n(x)$

 $\neq 0$ on I, and divide by it in order to rewrite the differential equation in the standard form

$$y^{(n)} + a_{n-1}y^{(n-1)} + \dots + a_1y' + a_0y = f.$$

 $(a_{n-1}, \dots a_1, a_0, f$ are all functions of x, and the DE above means that equality holds for all value of x in the interval I.)

This DE is called <u>linear</u> because the operator L defined by

$$L(y) := y^{(n)} + a_{n-1} y^{(n-1)} + \dots + a_1 y' + a_0 y$$

satisfies the so-called <u>linearity properties</u>

$$(1) L(y_1 + y_2) = L(y_1) + L(y_2)$$

$$(2) L(cy) = cL(y), c \in \mathbb{R}.$$

• The proof that L satisfies the linearity proporties is just the same as it was for the case when n = 2, that we checked Wednesday. Then, since the $y = y_p + y_H$ proof only depended on the linearity properties of L, *just like yesterday, we deduce both of Theorems 0 and 1:*

Theorem 0: The solution space to the homogeneous linear DE

$$y^{(n)} + a_{n-1}y^{(n-1)} + ... + a_1y' + a_0y = 0$$

is a subspace.

Theorem 1: The general solution to the <u>nonhomogeneous</u> n^{th} order linear DE $y^{(n)} + a_{n-1}y^{(n-1)} + ... + a_1y' + a_0y = f$

$$y^{(n)} + a_{n-1}y^{(n-1)} + ... + a_1y' + a_0y = f$$

is $y = y_P + y_H$ where y_P is any single particular solution and y_H is the general solution to the homogeneous DE. $(y_H \text{ is called } y_c, \text{ for complementary solution, in the text}).$

Later in the course we'll understand n^{th} order existence uniqueness theorems for initial value problems, in a way analogous to how we understood the first order theorem using slope fields, but let's postpone that discussion and just record the following true theorem as a fact:

Theorem 2 (Existence-Uniqueness Theorem): Let $a_{n-1}(x)$, $a_{n-2}(x)$,... $a_1(x)$, $a_0(x)$, f(x) be specified continuous functions on the interval I, and let $x_0 \in I$. Then there is a unique solution y(x) to the <u>initial value problem</u>

$$\begin{aligned} y^{(n)} + a_{n-1} y^{(n-1)} + \dots + a_1 y' + a_0 y = f \\ y(x_0) &= b_0 \\ y'(x_0) &= b_1 \\ y''(x_0) &= b_2 \\ &\vdots \\ y^{(n-1)}(x_0) &= b_{n-1} \end{aligned}$$

and y(x) exists and is n times continuously differentiable on the entire interval I.

Just as for the case n = 2, the existence-uniqueness theorem lets you figure out the dimension of the solution space to homogeneous linear differential equations. The proof is conceptually the same, but messier to write down because the vectors and matrices are bigger.

Theorem 3: The solution space to the n^{th} order homogeneous linear differential equation

$$y^{(n)} + a_{n-1} y^{(n-1)} + \dots + a_1 y' + a_0 y \equiv 0$$

is *n*-dimensional. Thus, any *n* independent solutions $y_1, y_2, \dots y_n$ will be a basis, and all homogeneous solutions will be uniquely expressible as linear combinations

$$y_H = c_1 y_1 + c_2 y_2 + \dots + c_n y_n.$$

<u>proof:</u> By the existence half of <u>Theorem 2</u>, we know that there are solutions for each possible initial value problem for this (homogeneous case) of the IVP for n^{th} order linear DEs. So, pick solutions $y_1(x), y_2(x), ..., y_n(x)$ so that their vectors of initial values (which we'll call initial value vectors)

$$\begin{bmatrix} y_1(x_0) \\ y_1{'}(x_0) \\ y_1{'}(x_0) \\ \vdots \\ y_1^{(n-1)}(x_0) \end{bmatrix}, \begin{bmatrix} y_2(x_0) \\ y_2{'}(x_0) \\ y_2{'}(x_0) \\ \vdots \\ y_2^{(n-1)}(x_0) \end{bmatrix}, \dots, \begin{bmatrix} y_n(x_0) \\ y_n{'}(x_0) \\ \vdots \\ y_n^{(n-1)}(x_0) \end{bmatrix}$$

are a basis for \mathbb{R}^n (i.e. these *n* vectors are linearly independent and span \mathbb{R}^n . (Well, you may not know how to "pick" such solutions, but you know they exist because of the existence theorem.)

<u>Claim:</u> In this case, the solutions $y_1, y_2, ... y_n$ are a basis for the solution space. In particular, every solution to the homogeneous DE is a unique linear combination of these n functions and the dimension of the solution space is n discussion on next page.

• Check that $y_1, y_2, \dots y_n$ **span** the solution space: Consider any solution y(x) to the DE. We can compute its vector of initial values

$$\begin{bmatrix} y(x_0) \\ y'(x_0) \\ y''(x_0) \\ \vdots \\ y^{(n-1)}(x_0) \end{bmatrix} := \begin{bmatrix} b_0 \\ b_1 \\ b_2 \\ \vdots \\ b_{n-1} \end{bmatrix}.$$

Now consider a linear combination $z = c_1 y_1 + c_2 y_2 + ... + c_n y_n$. Compute its initial value vector, and notice that you can write it as the product of the <u>Wronskian matrix</u> at x_0 times the vector of linear combination coefficients:

$$\begin{bmatrix} z(x_0) \\ z'(x_0) \\ \vdots \\ z^{(n-1)}(x_0) \end{bmatrix} = \begin{bmatrix} y_1(x_0) & y_2(x_0) & \dots & y_n(x_0) \\ y_1'(x_0) & y_2'(x_0) & \dots & y'_n(x_0) \\ \vdots & \vdots & \dots & \vdots \\ y_1^{(n-1)}(x_0) & y_2^{(n-1)}(x_0) & \dots & y_n^{(n-1)}(x_0) \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix}.$$

We've chosen the $y_1, y_2, ... y_n$ so that the Wronskian matrix at x_0 has an inverse, so the matrix equation

$$\begin{bmatrix} y_1(x_0) & y_2(x_0) & \dots & y_n(x_0) \\ y_1'(x_0) & y_2'(x_0) & \dots & y_n'(x_0) \\ \vdots & \vdots & \dots & \vdots \\ y_1^{(n-1)}(x_0) & y_2^{(n-1)}(x_0) & \dots & y_n^{(n-1)}(x_0) \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix} = \begin{bmatrix} b_0 \\ b_1 \\ \vdots \\ b_{n-1} \end{bmatrix}$$

has a unique solution \underline{c} . For this choice of linear combination coefficients, the solution $c_1y_1+c_2y_2+...+c_ny_n$ has the same initial value vector at x_0 as the solution y(x). By the uniqueness half of the existence-uniqueness theorem, we conclude that

$$y(x) = z(x) = c_1 y_1 + c_2 y_2 + \dots + c_n y_n$$
.

Thus $y_1, y_2, \dots y_n$ span the solution space.

- <u>linear independence</u>: If a linear combination $c_1y_1 + c_2y_2 + ... + c_ny_n \equiv 0$, then differentiate this identity n-1 times, and then substitute $x = x_0$ into the resulting n equations. This yields the Wronskian matrix equation above, with $\begin{bmatrix} b_0, b_1, ... b_{n-1} \end{bmatrix}^T = \begin{bmatrix} 0, 0, ..., 0 \end{bmatrix}^T$. So the matrix equation above implies that $\begin{bmatrix} c_1, c_2, ... c_n \end{bmatrix}^T = \mathbf{0}$. So $y_1, y_2, ... y_n$ are also <u>linearly independent</u>.
- Thus $y_1, y_2, \dots y_n$ are a <u>basis</u> for the solution space and the general solution to the homogeneous DE can be written as

$$y_H = c_1 y_1 + c_2 y_2 + \dots + c_n y_n$$
.

Let's do some new exercises that tie these ideas together. (We may do these excersises while or before we wade through the general discussions on the previous pages!

Exercise 1) Consider the 3^{rd} order linear homogeneous DE for y(x): y''' + 3y'' - y' - 3y = 0.

Find a basis for the 3-dimensional solution space, and the general solution. Make sure to use the Wronskian matrix (or determinant) to verify you have a basis. Hint: try exponential functions.

Exercise 2a) Find the general solution to

$$y''' + 3y'' - y' - 3y = 6$$
.

Hint: First try to find a particular solution ... try a constant function.

b) Set up the linear system to solve the initial value problem for this DE, with y(0) = -1, y'(0) = 2, y''(0) = 7.

for fun now, but maybe not just for fun later:

with (DEtools):

$$dsolve(\{y'''(x) + 3 \cdot y''(x) - y'(x) - 3 \cdot y(x) = 6, y(0) = -1, y'(0) = 2, y''(0) = 7\});$$

$$y(x) = -2 + \frac{9}{4} e^{x} + \frac{3}{4} e^{-3x} - 2 e^{-x}$$
(1)

• In section 5.2 there is a focus on testing whether collections of functions are linearly independent or not. This is important for finding bases for the solution spaces to homogeneous linear DE's because of the fact that if we find n linearly independent solutions to the n^{th} order homogeneous DE, they will automatically span the n-dimensional the solution space. (We discussed this general vector space "magic" fact on Wednesday.) And checking just linear independence is sometimes easier than also checking the spanning property.

Ways to check whether functions $y_1, y_2, \dots y_n$ are linearly independent on an interval:

In all cases you begin by writing the linear combination equation

$$c_1 y_1 + c_2 y_2 + \dots + c_n y_n = 0$$

where "0" is the zero function which equals 0 for all x on our interval.

<u>Method 1)</u> Plug in different x- values to get a system of algebraic equations for $c_1, c_2 ... c_n$. Either you'll get enough "different" equations to conclude that $c_1 = c_2 = ... = c_n = 0$, or you'll find a likely dependency.

Exercise 3) Use method 1 for $I = \mathbb{R}$, to show that the functions

$$y_1(x) = 1$$
, $y_2(x) = x$, $y_3(x) = x^2$

are linearly independent. (These functions show up in the homework due Monday.) For example, try the system you get by plugging in x = 0, -1, 1 into the equation

$$c_1 y_1 + c_2 y_2 + c_3 y_3 = 0$$

Method 2) If your interval stretches to $+\infty$ or to $-\infty$ and your functions grow at different rates, you may be able to take limits (after dividing the dependency equation by appropriate functions of x), to deduce independence.

Exercise 4) Use method 2 for $I = \mathbb{R}$, to show that the functions

$$y_1(x) = 1, \ y_2(x) = x, \ y_3(x) = x^2$$

are linearly independent. Hint: first divide the dependency equation by the fastest growing function, then let $x \to \infty$.

Method 3) If

$$c_1 y_1 + c_2 y_2 + \dots + c_n y_n = 0$$

 $\forall x \in I$, then we can take derivatives to get a system

$$c_{1}y_{1} + c_{2}y_{2} + \dots + c_{n}y_{n} = 0$$

$$c_{1}y_{1}' + c_{2}y_{2}' + \dots + c_{n}y_{n}' = 0$$

$$c_{1}y_{1}'' + c_{2}y_{2}'' + \dots + c_{n}y_{n}'' = 0$$

$$c_1 y_1^{(n-1)} + c_2 y_2^{(n-1)} + ... + c_n y_n^{(n-1)} = 0$$

(We could keep going, but stopping here gives us n equations in n unknowns.) Plugging in any value of x_0 yields a homogeneous algebraic linear system of n equations in n unknowns, which is equivalent to the Wronskian matrix equation

$$\begin{bmatrix} y_1(x_0) & y_2(x_0) & \dots & y_n(x_0) \\ y_1{'}(x_0) & y_2{'}(x_0) & \dots & y_n{'}(x_0) \\ \vdots & \vdots & \dots & \vdots \\ y_1^{(n-1)}(x_0) & y_2^{(n-1)}(x_0) & \dots & y_n^{(n-1)}(x_0) \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}.$$

If this Wronskian matrix is invertible at even a single point $x_0 \in I$, then the functions are linearly independent! (So if the determinant is <u>NON</u>-zero at even a single point $x_0 \in I$, then the functions are independent....strangely, even if the determinant was zero for all $x \in I$, then it could still be true that the functions are independent....but that won't happen if our n functions are all solutions to the same n^{th} order linear homogeneous DE.)

Exercise 5) Use method 3 for $I = \mathbb{R}$, to show that the functions

$$y_1(x) = 1$$
, $y_2(x) = x$, $y_3(x) = x^2$

are linearly independent. Use $x_0 = 1$.

<u>Remark 1)</u> Method 3 is usually <u>not</u> the easiest way to prove independence. But we and the text like it when studying differential equations because as we've seen, the Wronskian matrix shows up when you're trying to solve initial value problems using

$$y = y_P + y_H = y_P + c_1 y_1 + c_2 y_2 + \dots + c_n y_n$$

as the general solution to

$$y^{(n)} + a_{n-1}y^{(n-1)} + ... + a_1y' + a_0y = f.$$

This is because, if the initial conditions for this inhomogeneous DE are

$$y(x_0) = b_0, y'(x_0) = b_1, ... y^{(n-1)}(x_0) = b_{n-1}$$

then you need to solve matrix algebra problem

$$\begin{bmatrix} y_P(x_0) \\ y_{P^{'}}(x_0) \\ \vdots \\ y_P^{(n-1)}(x_0) \end{bmatrix} + \begin{bmatrix} y_1(x_0) & y_2(x_0) & \dots & y_n(x_0) \\ y_1^{'}(x_0) & y_2^{'}(x_0) & \dots & y_n^{'}(x_0) \\ \vdots & \vdots & \dots & \vdots \\ y_1^{(n-1)}(x_0) & y_2^{(n-1)}(x_0) & \dots & y_n^{(n-1)}(x_0) \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix} = \begin{bmatrix} b_0 \\ b_1 \\ \vdots \\ b_{n-1} \end{bmatrix}.$$

for the vector $[c_1, c_2, ..., c_n]^T$ of linear combination coefficients. And so if you're using the Wronskian matrix method, and the matrix is invertible at x_0 then you are effectively directly checking that $y_1, y_2, ..., y_n$ are a basis for the homogeneous solution space, and because you've found the Wronskian matrix you are ready to solve any initial value problem you want by solving for the linear combination coefficients above.

Remark 2) There is a seemingly magic consequence in the situation above, in which $y_1, y_2, \dots y_n$ are all solutions to the same n^{th} -order homogeneous DE

$$y^{(n)} + a_{n-1}y^{(n-1)} + ... + a_1y' + a_0y = 0$$

(even if the coefficients aren't constants): If the Wronskian matrix of your solutions $y_1, y_2, ..., y_n$ is invertible at a single point x_0 , then $y_1, y_2, ..., y_n$ are a basis because linear combinations uniquely solve all IVP's at x_0 . But since they're a basis, that also means that linear combinations of $y_1, y_2, ..., y_n$ solve all IVP's at any other point x_1 . This is only possible if the Wronskian matrix at x_1 also reduces to the identity matrix at x_1 and so is invertible there too. In other words, the Wronskian determinant will either be non-zero $\forall x \in I$, or zero $\forall x \in I$, when your functions $y_1, y_2, ..., y_n$ all happen to be solutions to the same n^{th} order homogeneous linear DE as above.

Exercise 6) Verify that $y_1(x) = 1$, $y_2(x) = x$, $y_3(x) = x^2$ all solve the third order linear homogeneous DE y''' = 0, and that their Wronskian determinant is indeed non-zero $\forall x \in \mathbb{R}$.