

Linear Algebra Notes

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MATH 3350 | Holt Linear Algebra 2nd Ed.

Ch 1

1.2 Intro stuff

Thm: Any system of linear eqs has either 0, exactly 1, or ∞ many solutions

How to: check how many solutions there are in a given system of eq:

If there last line is $0 = c$, then there is *no solution*

If not, are there free variables? If so, there are *inf many solutions*, otherwise *1 solution*.

Thm: Homogenous systems ALWAYS are consistent, aka they have either 1 or inf many solutions.

This is b/c the trivial solution will always exist when $A\mathbf{x} = 0$

Ch 2

2.1 Linear Combinations

How to: find all vectors $\begin{pmatrix} a \\ b \end{pmatrix}$ s.t. $c_1 u_1 + c_2 u_2 = \begin{pmatrix} a \\ b \end{pmatrix}$?

> Augment the vector with $\begin{pmatrix} a \\ b \end{pmatrix}$ and row reduce, the remaining things in the augment become the scalars c_1, c_2

How to: show that a certain vector \mathbf{b} cannot be obtained as a linear combination of some other vectors?

> Augment the matrix with \mathbf{b} then row reduce, you'll find an inconsistent set with $0 = c$

2.2 Spans

Definition of $\text{span}\{\mathbf{u}_1 \dots \mathbf{u}_m\}$ is the set of ALL linear combinations

How to: See if some vector \mathbf{v} is an element of $\text{span}\{u_1, u_2, \dots, u_m\}$?

> iff the linear system w/ \mathbf{v} as the augment has a solution

Thm: IF \mathbf{u}, \mathbf{v} are in $\text{span}\{\mathbf{u}_1 \dots \mathbf{u}_m\}$, THEN $\mathbf{u} + \mathbf{v}$ & $a\mathbf{u}$ are in that span
(linear combinations of the vectors are in that span)

How to: Find a vector \mathbf{b} not in a given span of vectors?

> Set matrix equal (by augment) to some (a, b, c) , and track what happens to it until row reduced.

> If the row is $[0 \ 0 \ 0 \mid f(c)]$, then any vector that has $f(c) \neq 0$ is not in the span

Thm: $(\text{span}\{u_1 \dots u_m\} = \mathbb{R}^n) \iff (B \text{ has a pivot in every row})$

How to: check if a set of vectors span \mathbb{R}^n ?

> Row reduce the matrix augmented with $[a \ b \ c]$

> Check if there is a row of 0s, which would imply that there could be a vector not in that span

> If there is a pivot in every row, there are either 1 or inf many ways to get to any point in \mathbb{R}^n
(depending on if there are free variables in any of the columns)

How to: Find what values of h in a $n \times m$ matrix allow it to span \mathbb{R}^n ?

a. Row reduce the matrix, moving the h down as needed to the last row.

b. The vectors in A span \mathbb{R}^n iff there is a trivial solution, which only occurs when $f(h) \neq 0$

c. The final answer should be all vectors with $\{h \mid h \neq 27\}$ or something

How to: find $\text{span}(\{\mathbf{a}_1 \dots \mathbf{a}_m\})$ (aka the columns of T)?

a.

Thm: For a given set of m vectors in \mathbb{R}^n :

a. IF $m < n$, THEN the set does not span \mathbb{R}^n (b/c theres no way to have a pivot in every row)

b. IF $m \geq n$, THEN the set could span \mathbb{R}^n (depends on whether or not they are linearly independent)

Thm: $\mathbf{b} \in \text{span}\{\mathbf{a}_1 \dots \mathbf{a}_m\} \text{ in } \mathbb{R}^n \iff A\mathbf{x} = \mathbf{b} \text{ has at least 1 solution}$

2.3 Linear Independence

Definition: IF the only way to express $\mathbf{0}$ as a linear combination of \mathbf{A} is the trivial solution $\mathbf{0}$, THEN the system is *linearly independent*. Nontrivial solutions imply *linear dependence*.

Thm: $(A\mathbf{x} = \mathbf{0} \text{ has only the trivial solution}) \iff (\{\mathbf{a}_1 \dots \mathbf{a}_m\} \text{ is linearly independent})$

How to: Check if a system is linearly independent?

> Set the system equal to $\mathbf{0}$, and row reduce. The only solution should be $\mathbf{x} = \mathbf{0}$ (which will happen when there is a pivot in every column)

Thm: $(m \text{ vectors in } \mathbb{R}^n \text{ are linearly independent}) \implies (m \leq n)$

(b/c you can't have pivots in every column if there are too many columns)
(in other words, there are more variables than equations in the system)

Thm: For a given set of m vectors in \mathbb{R}^n :

- $(\text{span}\{u_1 \dots u_m\} = \mathbb{R}^n) \iff (B \text{ has a pivot in every row})$
- $(\{u_1 \dots u_m\} \text{ are linearly independent}) \iff (B \text{ has a pivot in every column})$

How to: Check if one vector lies in the span of others in a set?

- Row reduce the matrix augmented with $\mathbf{0}$
- If B does not have a pivot in every column \implies the system is linearly dependent \implies one of the vectors lies in the span of the others

General vs. Particular Solutions to $A\mathbf{x}=\mathbf{0}$

Thm: $\mathbf{x}_g = \mathbf{x}_p + \mathbf{x}_h$, where \mathbf{x}_g is the solution to $A\mathbf{x} = \mathbf{b}$, \mathbf{x}_h is the solution to the associated homogenous system $A\mathbf{x} = \mathbf{0}$, and \mathbf{x}_p is a particular solution to \mathbf{x}_g

Thm: For a given set of vectors $\{\mathbf{a}_1 \dots \mathbf{a}_m\}$ and \mathbf{b} in \mathbb{R}^n :

- $(\{\mathbf{a}_1 \dots \mathbf{a}_m\} \text{ are linearly independent}) \iff (A\mathbf{x} = \mathbf{b})$

Ch 3

3.1 Linear Transformations

Definition: A transformation $T(\mathbf{x}) = A\mathbf{x}$ is linear if both:

- $T(\mathbf{u} + \mathbf{v}) = A(\mathbf{u} + \mathbf{v}) = A\mathbf{u} + A\mathbf{v} = T(\mathbf{u}) + T(\mathbf{v})$
- $T(r\mathbf{u}) = A(r\mathbf{u}) = rA\mathbf{u} = rT(\mathbf{u})$

Thm: $T(\mathbf{x}) = A\mathbf{x} \implies T$ is a linear transformation, where A is a $n \times m$ matrix, and T goes from \mathbb{R}^m to \mathbb{R}^n

How to: Check if a given transformation is linear:

- Convert the system into a matrix A
- Plug in the vectors $\begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$ for \mathbf{u} and $\begin{pmatrix} y_1 \\ y_2 \end{pmatrix}$ for \mathbf{v} to prove the general case *true*
- Try the basis vectors $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$ and $\begin{pmatrix} 0 \\ 1 \end{pmatrix}$ independently and check if the output fails to prove *false*

How to: Find $\text{range}(T)$, where $T(\mathbf{x}) = A\mathbf{x}$:

- $\text{range}(T) = \text{span}(\{\mathbf{a}_1 \dots \mathbf{a}_m\})$
- range is the set of linear combinations of the columns of A

How to: Check if a given vector \mathbf{w} is in $\text{range}(T)$:

- Make matrix of $[A \mid \mathbf{w}]$ and solve

One-to-One vs Onto

Definition: A transformation is *one-to-one* when there's at most one input that maps to an output

Definition: A transformation is *onto* when no element in the codomain B is left out

Thm: Given $T : \mathbb{R}^m \rightarrow \mathbb{R}^n$ and $T(\mathbf{x}) = \mathbf{A}\mathbf{x}$, where \mathbf{B} is \mathbf{A} in row-echelon form:

1a. (T is one-to-one) \iff (columns of \mathbf{A} are linearly independent) \iff (\mathbf{B} has a pivot in every column)

1b. $n < m \implies T$ is not one-to-one (aka if output space is smaller than input space)

2a. (T is onto) \iff (columns of \mathbf{A} span the codomain \mathbb{R}^n aka $\text{range}(T) = \mathbb{R}^n$) \iff (\mathbf{B} has a pivot in every row)

2b. $n > m \implies T$ is not onto (aka if output space is bigger than input space)

> "No matrix that goes from bigger space to smaller space can be one-to-one"

> "No matrix that goes from small space to bigger space can be onto"

Geometry of transformations

How to: Rotate a vector CCW by θ :

a. $T_r(\mathbf{x}) = \begin{pmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{pmatrix} \mathbf{x}$

How to: Shear to the right:

a. $T_r(\mathbf{x}) = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} \mathbf{x}$

3.2 Matrix Algebra

Properties of Elementary Matrices

a. $A(BC) = (AB)C$

b. $A(B + C) = AB + AC$

c. $(A + B)C = AC + BC$

d. $s(AB) = (sA)B = A(sB)$

e. $AI = IA = A$

Non-Properties of Nonzero Matrices

a. It is possible that $AB \neq BA$

b. $AB = 0$ does not imply that $A = 0$ and $B = 0$

c. $AC = BC$ does not imply that $A = B$ or $C = 0$ (unless A is invertible)

Transpose of a Matrix

- a. $(A + B)^T = A^T + B^T$
- b. $(sA)^T = sA^T$
- c. $(AC)^T = C^T A^T$

3.3 Inverses

Definition: If T is a linear transformation, Then

- a. T has an inverse $\implies m = n$
- b. If T is invertible, then T^{-1} is also a linear transformation

$(T \text{ is invertible}) \iff (T \text{ is one-to-one AND onto})$

How to: find an invertible matrix A^{-1} ?

- a. Augment matrix A with I_n , then row reduce until you get I_n augmented with A^{-1}
(aka $[A|I_n] \rightarrow [I_n|A^{-1}]$)

Thm: Elementary matrices are invertible

Properties of Inverses

- a. $(A^{-1})^{-1} = A$
- b. $(AB)^{-1} = B^{-1}A^{-1}$
- c. $AC = AD \implies C = D$
- d. $AC = 0_{nm} \implies C = 0_{nm}$

Ch 4

4.1 Subspaces

Definition: A subset of S is a subspace if all three conditions are true:

- a. S contains $\mathbf{0}$ (S contains the origin)
- b. If \mathbf{u} and \mathbf{v} are both in S , then $(\mathbf{u} + \mathbf{v})$ is in S (S is closed under addition)
- c. If $r \in \mathbb{R}$, then $r\mathbf{u}$ is also in S (S is closed under scalar multiplication)

Thm: If $S = \text{span}\{\mathbf{u}_1 \dots \mathbf{u}_m\}$ in \mathbb{R}^n , then S is a subspace of \mathbb{R}^n

How to: Check if S is a subspace?

- a. Check if $\mathbf{0}$ is in S , which it must be to be a subspace
- b. Try to show S is generated by a set of vectors (See if it can be composed as a matrix of coefficients)

Definition: If \mathbf{A} is a $n \times m$ matrix, then the set of solutions to $\mathbf{Ax} = \mathbf{0}$ is called $\text{null}(\mathbf{A})$

(aka the null space is all linear combinations where $\mathbf{Ax} = \mathbf{0}$)

(aka the null space is the solution to the homogenous system)

Thm: If \mathbf{A} is a $n \times m$ matrix, then the set of solutions to $\mathbf{Ax} = \mathbf{0}$ forms a subspace of \mathbb{R}^m
(aka null space is a subspace)

Thm: Given $T : \mathbb{R}^m \rightarrow \mathbb{R}^n$ is a *linear* transformation:

- $\ker(T)$ is a subspace of the domain \mathbb{R}^m
- $\text{range}(T)$ is a subspace of the codomain \mathbb{R}^n

> "The kernel is the set of vectors that are sent to $\{\mathbf{0}\}$ after applying T "

> "The range of T is the span after applying T "

How to: Find $\ker(T)$ of $T(\mathbf{x}) = A(\mathbf{x})$?

a. $(T(\mathbf{x}) = \mathbf{Ax}) \implies (\ker(T) = \text{null}(A))$, so solve for $\mathbf{Ax} = \mathbf{0}$, and $\ker(T)$ is the span of that answer

How to: Find $\text{null}(T)$ of $T(\mathbf{x}) = A(\mathbf{x})$?

a. $\text{range}(T) = \text{span}(\mathbf{a}_1 \dots \mathbf{a}_m)$, so just delete any linearly dependent columns of \mathbf{A} and that's your answer

Thm: (T is one-to-one) $\iff (\ker(T) = \{\mathbf{0}\})$

4.2 Basis vectors

Definition: Set $B = \{\mathbf{u}_1 \dots \mathbf{u}_m\}$ is a *basis* of subspace S iff:

- B spans S
- B is linearly independent

"To get to any point in S , you can take a linear combination of the basis vectors to get there"

How to: find a basis for $S = \text{span}\{\mathbf{u}_1 \dots \mathbf{u}_m\}$?

Method 1 (Thm 4.10):

- Create a matrix $\begin{pmatrix} \mathbf{u}_1 \\ \dots \\ \mathbf{u}_m \end{pmatrix}$
- Row reduce to B
- The nonzero rows of B give a basis of S

Method 2 (Thm 4.11):

- Create a matrix out of $\{\mathbf{u}_1 \dots \mathbf{u}_m\}$
- Row reduce to B . The pivot columns of B are linearly independent
(the other cols will depend on the pivot columns)
- The columns of A corresponding to the pivot columns of B form a basis of S .

Dimension

Thm: If S is a subspace of \mathbb{R}^n , then every basis of S has the same number of vectors

Definition: If S is a subspace of \mathbb{R}^n , then the dimension of S is the number of vectors in any basis of S

Thm: If $U = \{\mathbf{u}_1 \dots \mathbf{u}_m\}$ is a set of m vectors in subspace S of dimension m , if U is *either* linearly independent or spans S , THEN U is a basis for S .

Unifying Theorem: Given $S = \{a_1 \dots a_m\}, \{a_1 \dots a_m\} \in \mathbb{R}^n, A = [\mathbf{a}_1 \dots \mathbf{a}_m]$, and $T: \mathbb{R}^m \rightarrow \mathbb{R}^n, T(\mathbf{x}) = A\mathbf{x}$:

- S spans \mathbb{R}^n
- S is linearly independent
- $A\mathbf{x} = \mathbf{b}$ has precisely 1 unique solution $\forall \mathbf{b} \in \mathbb{R}^n$
- T is onto
- T is one-to-one
- A is invertible
- $\ker\{T\} = \{\mathbf{0}\} \iff \text{null}(A) = \{\mathbf{0}\}$
- S is a basis of \mathbb{R}^n

Ch 4.3 Row and Column Spaces

Definition: Given A is a $n \times m$ matrix:

- $\text{row}(A)$ or row space is the subspace of \mathbb{R}^m spanned by row vectors of A
- $\text{col}(A)$ or column space is the subspace of \mathbb{R}^n spanned by column vectors of A

How to:

Thm: Given matrix A and B in echelon form:

- Nonzero rows of B form a basis for $\text{row}(A)$
- The cols of A corresponding to pivot columns of B form a basis for $\text{col}(A)$

Thm: For any matrix A , the dimension of $\text{row}(A)$ equals the dimension of $\text{col}(A)$

Definition: $\text{rank}(A)$ is the dimension of $\text{row}(A)$ or $\text{col}(A)$

Rank-Nullity Thm: IF A is a $n \times m$ matrix, THEN $\text{rank}(A) + \text{nullity}(A) = m$

*Note: $\ker(T)$ is $\text{null}(A)$, and $\text{range}(T)$ is $\text{col}(A)$

Thm: IF A is a $n \times m$ matrix, and \mathbf{b} is a vector in \mathbb{R}^n :

- The system $A\mathbf{x} = \mathbf{b}$ is consistent $\iff \mathbf{b}$ is in $\text{col}(A)$
- The system $A\mathbf{x} = \mathbf{b}$ has a unique solution $\iff \mathbf{b}$ is in $\text{col}(A)$ and columns of A are linearly independent

Unifying Theorem: Given $S = \{a_1 \dots a_m\}, \{a_1 \dots a_m\} \in \mathbb{R}^n, A = [\mathbf{a}_1 \dots \mathbf{a}_m]$, and $T: \mathbb{R}^m \rightarrow \mathbb{R}^n, T(\mathbf{x}) = A\mathbf{x}$:

- S spans \mathbb{R}^n
- S is linearly independent

- c. $A\mathbf{x} = \mathbf{b}$ has precisely 1 unique solution $\forall \mathbf{b} \in \mathbb{R}^n$
- d. T is onto
- e. T is one-to-one
- f. A is invertible
- g. $\ker\{T\} = \{\mathbf{0}\} \iff \text{null}(A) = \{\mathbf{0}\}$
- h. S is a basis of \mathbb{R}^n
- i. $\text{col}(A) = \mathbb{R}^n$
- j. $\text{col}(A) = \mathbb{R}^n$
- k. $\text{rank}(A) = n$

Ch 4.4 Change of Basis

Definition: Suppose that $B = \{\mathbf{u}_1 \dots \mathbf{u}_n\}$ forms a basis of \mathbb{R}^n , and if $\mathbf{y} = y_1 \mathbf{u}_1 + \dots + y_n \mathbf{u}_n$:

THEN: The *coordinate vector of y w.r.t. B is $[\mathbf{y}]_B = \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}$

"the coordinate vector contains the coeffs required to express y as a linear combination of vectors in basis B "

$$U \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix} = y_1 \mathbf{u}_1 + \dots + y_n \mathbf{u}_n$$

$\mathbf{y} = U[\mathbf{y}]_B$; where U is the *change of basis matrix* that transforms the coordinate vector wrt B back to the standard basis

(U is just the $n \times n$ matrix containing the basis vectors of set B : $[\mathbf{u}_1 \dots \mathbf{u}_n]$)

Thm: Let \mathbf{x} be expressed wrt standard basis, and $B = \{\mathbf{u}_1 \dots \mathbf{u}_n\}$ be any basis for \mathbb{R}^n :

If $U = [\mathbf{u}_1 \dots \mathbf{u}_n]$, then: $\mathbf{x} = U[\mathbf{x}]_B$ and $[\mathbf{x}]_B = U^{-1}\mathbf{x}$

How to: Move from one nonstandard basis to another?

If $B_1 = \{\mathbf{u}_1 \dots \mathbf{u}_n\}$ corresponds to U and $B_2 = \{\mathbf{v}_1 \dots \mathbf{v}_n\}$ corresponds to V , then:

$$[\mathbf{x}]_{B_2} = V^{-1}U[\mathbf{x}]_{B_1}$$

$$[\mathbf{x}]_{B_1} = U^{-1}V[\mathbf{x}]_{B_2}$$

> "To go from basis 1 to basis 2, apply U to go into standard basis, then apply V^{-1} to return to basis 2 land"

