## March 16 Math 3260 sec. 51 Spring 2020

## Section 4.4: Coordinate Systems

Definition: (Coordinate Vectors) Let $\mathcal{B}=\left\{\mathbf{b}_{1}, \ldots, \mathbf{b}_{n}\right\}$ be an ordered basis of the vector space $V$. For each $\mathbf{x}$ in $V$ we define the coordinate vector of $\mathbf{x}$ relative to the basis $\mathcal{B}$ to be the unique vector $\left(c_{1}, \ldots, c_{n}\right)$ in $\mathbb{R}^{n}$ where these entries are the weights $\mathbf{x}=c_{1} \mathbf{b}_{1}+\cdots c_{n} \mathbf{b}_{n}$.

We'll use the notation

$$
\left[\begin{array}{c}
c_{1} \\
c_{2} \\
\vdots \\
c_{n}
\end{array}\right]=[\mathbf{x}]_{\mathcal{B}} .
$$

## Coordinates in $\mathbb{R}^{n}$

Change of Coordinates in $\mathbb{R}^{n}$ : Let $\mathcal{B}=\left\{\mathbf{b}_{1}, \ldots, \mathbf{b}_{n}\right\}$ be an ordered basis of $\mathbb{R}^{n}$. Then the change of coordinate mapping $\mathbf{x} \mapsto[\mathbf{x}]_{\mathcal{B}}$ is the linear transformation defined by

$$
[\mathbf{x}]_{\mathcal{B}}=P_{\mathcal{B}}^{-1} \mathbf{x}
$$

where the matrix

$$
P_{\mathcal{B}}=\left[\begin{array}{llll}
\mathbf{b}_{1} & \mathbf{b}_{2} & \cdots & \mathbf{b}_{n}
\end{array}\right] .
$$

## Theorem: Coordinate Mapping

Let $\mathcal{B}=\left\{\mathbf{b}_{1}, \ldots, \mathbf{b}_{n}\right\}$ be an ordered basis for a vector space $V$. Then the coordinate mapping $\mathbf{x} \mapsto[\mathbf{x}]_{\mathcal{B}}$ is a one to one mapping of $V$ onto $\mathbb{R}^{n}$.

Remark: When such a mapping exists, we say that $V$ is isomorphic to $\mathbb{R}^{n}$. Properties of subsets of $V$, such as linear dependence, can be discerned from the coordinate vectors in $\mathbb{R}^{n}$.

## Example

Use coordinate vectors to determine if the set $\{\mathbf{p}, \mathbf{q}, \mathbf{r}\}$ is linearly dependent or independent in $\mathbb{P}_{2}$.

$$
\mathbf{p}(t)=1-2 t^{2}, \quad \mathbf{q}(t)=3 t+t^{2}, \quad \mathbf{r}(t)=1+t
$$

## Section 4.5: Dimension of a Vector Space

Theorem: If a vector space $V$ has a basis $\mathcal{B}=\left\{\mathbf{b}_{1}, \ldots, \mathbf{b}_{n}\right\}$, then any set of vectors in $V$ containing more than $n$ vectors is linearly dependent.

We saw an example of this before: a set of $p$ vectors in $\mathbb{R}^{n}$ is linearly dependent if $p>n$.

Why is the set $\{1+t, 1-2 t, 2+4 t\}$ linearly dependent in $\mathbb{P}_{1}$ ?

## Dimension

Corollary: If vector space $V$ has a basis $\mathcal{B}=\left\{\mathbf{b}_{1}, \ldots, \mathbf{b}_{n}\right\}$, then every basis of $V$ consist of exactly $n$ vectors.

- What this says is that the number of basis elements for a given vector space is fixed.
- This number can be used, unambiguously, as a characteristic of the vector space. This leads to the definition of dimension.


## Dimension of a Vector Space

Definition: If $V$ is spanned by a finite set, then $V$ is called finite dimensional.

In this case, the dimension of $V$

$$
\operatorname{dim} V=\text { the number of vectors in any basis of } V
$$

The dimension of the vector space $\{\mathbf{0}\}$ containing only the zero vector is defined to be zero-i.e.

$$
\operatorname{dim}\{\mathbf{0}\}=0
$$

If $V$ is not spanned by a finite set ${ }^{1}$, then $V$ is said to be infinite dimensional.
${ }^{1} C^{0}(\mathbb{R})$ is an example of an infinite dimensional vector space.

## Examples

(a) Find $\operatorname{dim}\left(\mathbb{R}^{n}\right)$.

## Examples

(b) Determine $\operatorname{dim}$ Col $A$ where $A=\left[\begin{array}{ccc}1 & 1 & 3 \\ 0 & 0 & -1\end{array}\right]$.

## Some Geometry in $\mathbb{R}^{3}$

The subspaces of $\mathbb{R}^{3}$ of various dimensions are:

- Zero: One point, the origin.
- One: Any line through the origin, e.g. $\operatorname{Span}\{\mathbf{u}\}$.
- Two: Any plane that contains the origin, e.g. $\operatorname{Span}\{\mathbf{u}, \mathbf{v}\}$.
- Three: All of $\mathbb{R}^{3}$

Note: It is assumed that vectors $\mathbf{u}$ and $\mathbf{v}$ are linearly independent, nonzero vectors.

## Subspaces and Dimension

Theorem: Let $H$ be a subspace of a finite dimensional vector space $V$. Then $H$ is finite dimensional and

$$
\operatorname{dim} H \leq \operatorname{dim} V
$$

Moreover, any linearly independent subset of $H$ can be expanded if needed to form a basis for $H$.

We already knew that if we had a spanning set, we could obtain a basis (by getting rid of duplicating vectors).

This says if we start with a linearly independent set, we can add linearly independent vectors as needed, until we get a spanning set.

## Subspaces and Dimension

Theorem: Let $V$ be a vector space with $\operatorname{dim} V=p$ where $p \geq 1$. Any linearly independent set in $V$ containing exactly $p$ vectors is a basis for $V$. Similarly, any spanning set consisting of exactly $p$ vectors in $V$ is necessarily a basis for $V$.

That is: If a set of $p$ vectors in a $p$-dimensional vectors space $V$
(1) spans $V$, it is automatically linearly independent.
(2) is linearly independent, it automatically spans $V$.

## Column and Null Spaces

Theorem: Let $A$ be an $m \times n$ matrix. Then $\operatorname{dim} \operatorname{Nul} A=$ the number of free variables in the equation $A \mathbf{x}=\mathbf{0}$,
and
$\operatorname{dim} \operatorname{Col} A=$ the number of pivot positions in $A$.

## Example

Find the dimensions of the null and columns spaces of the matrix $A$.

$$
A=\left[\begin{array}{cccc}
1 & 1 & 1 & 0 \\
-3 & 1 & -7 & -1 \\
3 & 0 & 6 & 1
\end{array}\right]
$$



