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Math 217 Exam Two Notes

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COORDINATES

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Coordinates Matrices

Book Theorem 4.3.2

The \mathfrak{B} -matrix of T

$$[T]_{\mathfrak{B}} = \begin{bmatrix} | & | & | \\ [T(\vec{v}_1)]_{\mathfrak{B}} & [T(\vec{v}_2)]_{\mathfrak{B}} & \cdots & [T(\vec{v}_n)]_{\mathfrak{B}} \\ | & | & | \end{bmatrix}$$
$$[T]_{\mathfrak{B}} = S_{\mathfrak{U} \to \mathfrak{B}}[T]_{\mathfrak{U}}S_{\mathfrak{B} \to \mathfrak{U}}$$

Book Theorem 4.3.2

Change of Basis Matrix $S_{\mathfrak{U} \to \mathfrak{B}}$

$$S_{\mathfrak{U}\to\mathfrak{B}} = (S_{\mathfrak{B}\to\mathfrak{U}})^{-1}$$

$$S_{\mathfrak{U}\to\mathfrak{B}} = \begin{bmatrix} | & | & | \\ [\vec{u}_1]_{\mathfrak{B}} & [\vec{u}_2]_{\mathfrak{B}} & \cdots & [\vec{u}_n]_{\mathfrak{B}} \\ | & | & | \end{bmatrix}$$

Book Theorem 4.3.4

Change of Basis in a subspace of \mathbb{R}^n

$$\begin{bmatrix} \begin{vmatrix} & & & & \\ \vec{b}_1 & \vec{b}_2 & \cdots & \vec{b}_n \\ & & & \end{vmatrix} = \begin{bmatrix} \begin{vmatrix} & & & & \\ \vec{u}_1 & \vec{u}_2 & \cdots & \vec{u}_n \\ & & & & \end{vmatrix} S_{\mathfrak{B} \to \mathfrak{U}}$$

Coordinate Theorems

Worksheet 16 Problem 3

Suppose that $\mathfrak{U}=(\vec{u}_1,\vec{u}_2,...,\vec{u}_n)$ is an orthonormal ordered basis for \mathbb{R}^n

$$[\vec{x}]_{\mathfrak{U}} = \begin{bmatrix} \vec{x} \cdot \vec{u}_1 \\ \vec{x} \cdot \vec{u}_2 \\ \vdots \\ \vec{x} \cdot \vec{u}_n \end{bmatrix}$$

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ORTHOGONALITY

Orthogonal Complement Theorems

Book Theorem 5.4.1

$$(\operatorname{im} A)^{\perp} = \ker(A^{T})$$
$$\operatorname{im}(A^{T}) = (\ker A)^{\perp}$$

Book Theorem 5.1.8

If V is a subspace, then

$$V^{\perp}$$
 is a subspace
$$(V^{\perp})^{\perp} = V$$

$$n = \dim(V) + \dim(V^{\perp})$$

$$V \cap V^{\perp} = \vec{0}$$

Homework 7 Problem 4

If V is a subspace, then

$$V^{\perp} + W^{\perp} = (V \cap W)^{\perp}$$
$$(V + W)^{\perp} = V^{\perp} \cap W^{\perp}$$

Orthogonal Matrix Characterizations

Book Theorem 5.3.8

$$A^{T}A = I_{n}$$

$$A^{-1} = A^{T}$$

$$||A\vec{x}|| = ||\vec{x}||$$

$$(A\vec{x}) \cdot (A\vec{y}) = \vec{x} \cdot \vec{y}$$

Book Theorem 5.3.4

if A and B are orthogonal matrices

AB is orthogonal A^{-1} is orthogonal

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Graham Schmidt

Book Theorem 5.2.1

Construct an orthonormal basis $(\vec{u}_1,...,\vec{u}_n)$ from any basis $(\vec{v}_1,...,\vec{v}_n)$

$$ec{u}_j = rac{ec{v}_j^{\perp}}{||ec{v}_j^{\perp}||} \ ec{v}_j^{\perp} = ec{v}_j - (ec{u}_1 \cdot ec{v}_j) ec{u}_1 - \dots - (ec{u}_{j-1} \cdot ec{v}_j j) ec{u}_{j-1}$$

QR Factorization

Book Theorem 5.2.2

QR factorization of a matrix M

DOT PRODUCT & PROJECTIONS

Dot Product Theorems

Book Theorem 5.1.1

$$\begin{split} ||\vec{v}|| &= \sqrt{\vec{v} \cdot \vec{v}} \\ \vec{v} \cdot \vec{w} &= 0 \iff \vec{v} \perp \vec{w} \\ \vec{u} \cdot \vec{u} &= 1 \iff \vec{u} \text{ is a unit vector} \end{split}$$

Book Theorem 5.3.6

$$A\vec{v}\cdot\vec{w} = \vec{v}\cdot A^T\vec{w}$$

Book Theorem 5.1.9 - 5.1.12

$$||\vec{x} + \vec{y}||^2 = ||\vec{x}||^2 + ||\vec{y}||^2$$
 Pythagorean Theorem *for orthogonal vectors $||\operatorname{proj}_V \vec{x}|| \leq ||\vec{x}||$ Magnitude of $\operatorname{proj}_V \vec{x}$ $|\vec{x} \cdot \vec{y}| \leq ||\vec{x}|| \ ||\vec{y}||$ Cauchy-Schwarz Inequality

 $\cos \theta = \frac{\vec{x} \cdot \vec{y}}{||\vec{x}|| \ ||\vec{y}||}$ Angle between two vecors

Projections

Book Theorem 5.1.5

If V is a subspace of \mathbb{R}^n with orthonormal basis $\mathfrak{U} = (\vec{u}_1, \vec{u}_2, ..., \vec{u}_n)$ then

$$\operatorname{proj}_{V} \vec{x} = (\vec{u}_{1} \cdot \vec{x}) \vec{u}_{1} + (\vec{u}_{2} \cdot \vec{x}) \vec{u}_{2} + \dots + (\vec{u}_{n} \cdot \vec{x}) \vec{u}_{n}$$

for all $\vec{x} \in \mathbb{R}^n$

Orthogonal Projection

Book Theorem 5.1.5

If V is a subspec of \mathbb{R}^n with orthogonal basis $\mathfrak{B} = (\vec{v}_1, \vec{v}_2, ..., \vec{v}_n)$ then $\operatorname{proj}_V \vec{x} = \frac{\vec{v}_1 \cdot \vec{x}}{\vec{v}_1 \cdot \vec{v}_1} \vec{v}_1 + \frac{\vec{v}_2 \cdot \vec{x}}{\vec{v}_2 \cdot \vec{v}_2} \vec{v}_2 + \cdots + \frac{\vec{v}_n \cdot \vec{x}}{\vec{v}_n \cdot \vec{v}_n} \vec{v}_n$

for all $\vec{x} \in \mathbb{R}^n$

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Matrix of an Orthogonal Projection

Book Theorem 4.3.4

Projection onto a subspace $V \subseteq \mathbb{R}^n$ with orthonormal basis $\mathfrak{U} = (\vec{u}_1, \vec{u}_2, ..., \vec{u}_d)$ and basis $\mathfrak{B} = (\vec{b}_1, \vec{b}_2, ..., \vec{b}_d)$

$$P = QQ^{T} \quad \text{where} \quad Q \qquad \qquad = \begin{bmatrix} | & | & & | \\ \vec{u}_{1} & \vec{u}_{2} & \cdots & \vec{u}_{d} \\ | & | & & | \end{bmatrix}$$

$$P = A(A^{T}A)^{-1}A^{T} \quad \text{where} \quad A \quad = \begin{bmatrix} | & | & & | \\ \vec{b}_{1} & \vec{b}_{2} & \cdots & \vec{b}_{d} \\ | & | & & | \end{bmatrix}$$

Book Theorem 5.4.6

 $A\vec{x} = \vec{b}$ has the unique solution

$$\vec{x}^* = (A^T A)^{-1} A^T \vec{b}$$

TRANSPOSE & LEAST SQUARES

Transpose Theorems

Book Theorem 5.3.9

$$(A+B)^{T} = A^{T} + B^{T}$$
$$(kA)^{T} = kA^{T}$$
$$(AB)^{T} = B^{T}A^{T}$$
$$\operatorname{rank}(A) = \operatorname{rank}(A^{T})$$
$$(A^{T})^{-1} = (A^{-1})^{T}$$

Homework 8 Problem 3

$$\ker(A) = \ker(A^T A)$$
$$\operatorname{im}(A) = \operatorname{im}(AA^T)$$

Least Squares Definition

Book Theorem 5.4.5

 $A\vec{x} = \vec{b}$ has the least-squares solutions of the consistent system

$$A^T A \vec{x} = A^T \vec{b}$$

Book Theorem 5.4.6

If $ker(A) = {\vec{0}}$, then linear system $A\vec{x} = \vec{b}$ has the unique solution

$$\vec{x}^* = (A^T A)^{-1} A^T \vec{b}$$

Least Squares Theorems

Book Theorem 5.4.6

If $ker(A) = {\vec{0}}$, then linear system $A\vec{x} = \vec{b}$ has the unique solution

$$\vec{x}^* = (A^T A)^{-1} A^T \vec{b}$$

GEOMETRIC TRANSFORMATIONS

Projections

Rotations

Reflections