

**Catalog description:** A study of vectors, matrices and linear transformations, principally in two and three dimensions, including treatments of systems of linear equations, determinants, and eigenvalues. Prerequisite: MAT 1120 or permission of the instructor.

### Course Goals

- Develop algebraic skills
- Develop mathematical reasoning and problem solving
- Develop spatial visualization skills
- Learn about some applications of linear algebra
- An introduction to a computer algebra software system as it applies to linear algebra

### Mapping of the Topics in the Catalog Description to the Text

Systems of Linear Equations: 1.1, 1.2, 1.5

Vectors: 1.3, 1.4, 1.7, 6.1

Matrices: earlier +2.1, 2.2, 2.3

Linear transformations 1.8, 1.9 (1-1 and onto eliminated), 2.7

Determinants: 3.1, 3.2, 3.3

Eigenvalues 2.8, 5.1, 5.2, 5.6

## Test 2: cumulative + what we covered in 1.8, 1.9, 2.7, 2.8, 3.1, 3.2, 3.3, 5.1, 5.6, 6.1 & apps

- Formatting same as prior tests. Test 2 is majority new material.
- hw, problem sets, & clicker questions [solutions online]
- computations, definitions, critical reasoning & “big picture”
- 1.1, 1.2 & 1.5: Gaussian elimination, algebra and geometry of solutions of systems of equations...
- 1.4: connects everything together
- 1.3 and 1.7: algebra and geometry of vectors (linear combinations/mixing, span, li...)
- 2.1 and 2.2: matrix algebra:  $A + B$ ,  $cA$ ,  $A^T$ ,  $AB$ ,  $A_{2 \times 2}^{-1}$ ,  $\det(A_{2 \times 2})$
- 2.3: theorem 8: what makes a matrix invertible [connects 2.2 to 1.1, 1.2, 1.3 and 1.7] & condition #
- 1.8 (62, 65, 67-68), 1.9 (70-75), 2.7: linear transformations
- apps: Hill cipher, computer graphics/animations, computer speed and reliability
- 3.1-3.3: algebra and geometry of determinants, invertibility
- 2.8: subspace, basis, column space and null space
- 5.1: eigenvectors & eigenvalues alg & geom,  $\text{nullspace}(A - \lambda I)$
- 5.6: eigenvector decomposition, limit, trajectory & populations
- 6.1 : length & angle of a vector, orthogonal vectors

$$\text{Rotation: } \begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix} \quad \text{Dilation: } \begin{bmatrix} c & 0 \\ 0 & c \end{bmatrix} \quad \text{Horizontal Shear: } \begin{bmatrix} 1 & k \\ 0 & 1 \end{bmatrix}$$

$$\text{Projections: } y=x \text{ line: } \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix} \quad x\text{-axis: } \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \quad y\text{-axis: } \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\text{Reflections: } y=x \text{ line: } \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \quad x\text{-axis: } \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} \quad y\text{-axis: } \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\text{Translation: } \begin{bmatrix} 1 & 0 & h \\ 0 & 1 & k \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} x+h \\ y+k \\ 1 \end{bmatrix} \quad \text{Others: } \begin{bmatrix} a & b & 0 \\ c & d & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

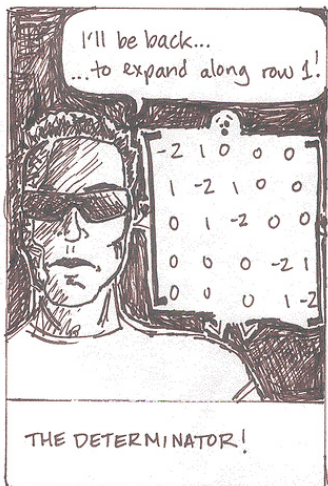
Rotate a Figure about the point  $\begin{bmatrix} 4 \\ 9 \end{bmatrix}$  :

$$\begin{bmatrix} 1 & 0 & 4 \\ 0 & 1 & 9 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \cos(\theta) & -\sin(\theta) & 0 \\ \sin(\theta) & \cos(\theta) & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & -4 \\ 0 & 1 & -9 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 & \dots & x_p \\ y_1 & \dots & y_p \\ 1 & \dots & 1 \end{bmatrix}$$

$$\begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix} \text{ with columns: } \vec{u} = \begin{bmatrix} \cos(\theta) \\ \sin(\theta) \end{bmatrix}, \vec{v} = \begin{bmatrix} -\sin(\theta) \\ \cos(\theta) \end{bmatrix}$$

Length (or norm) of  $\vec{u} = \|\vec{u}\| \sqrt{\vec{u} \cdot \vec{u}} = \sqrt{\begin{bmatrix} \cos(\theta) \\ \sin(\theta) \end{bmatrix} \cdot \begin{bmatrix} \cos(\theta) \\ \sin(\theta) \end{bmatrix}}$

- Useful in keeping a car on a racetrack. If we don't normalize the vectors, the car size won't be preserved.
- Here **inner product**  $\vec{u} \cdot \vec{v} = \vec{u}^T \vec{v}$  induces metric on the space  $\|\vec{u} - \vec{v}\|$  is distance between vectors as in 1.3
- Generalized inner products for nonlinear/non-Euclidean satisfy axiomatic properties like distributivity, pulling out scalars, positive definite condition
- Two vectors are **orthogonal** if right angle between them. One formulation of the dot product  $\vec{u} \cdot \vec{v}$  is  $\|\vec{u}\| \|\vec{v}\| \cos\theta$ , where  $\theta$  is the angle between them, so the dot product is 0 exactly when the angle is  $\frac{\pi}{2}$

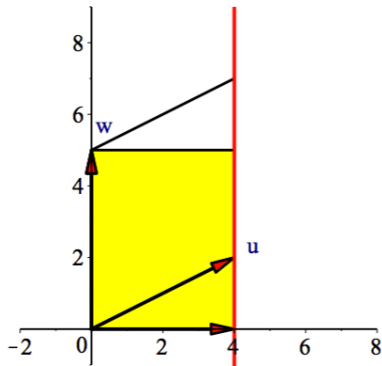


2007

@COURTNEY GIBBONS

$$-2 \cdot (-1)^{1+1} \begin{vmatrix} -2 & 1 & 0 & 0 \\ 1 & -2 & 0 & 0 \\ 0 & 0 & -2 & 1 \\ 0 & 0 & 1 & -2 \end{vmatrix} + 1 \cdot (-1)^{1+2} \begin{vmatrix} 1 & 1 & 0 & 0 \\ 0 & -2 & 0 & 0 \\ 0 & 0 & -2 & 1 \\ 0 & 0 & 1 & -2 \end{vmatrix} + 0s$$





Replacement  $r'_j = -3r_1 + r_j$  are shear matrices when written in elementary matrix form like  $\begin{bmatrix} 1 & 0 \\ -3 & 1 \end{bmatrix}$  or  $\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -3 & 0 & 1 \end{bmatrix}$ —preserve the determinant (area, volume...) of a matrix and turns parallelograms and parallelepipeds formed by the vectors into rectangles and rectangular prisms

## What Makes You Invertible

In **verse**

Music by One Direction & idea adapted from Art Benjamin  
Interpreted by Dr. Sarah and Joel Landsberg

Baby you'll light up if one of these facts is so,  
but you'll need  $n$  square columns and rows:

- Like when  $\mathbb{R}^n$  is the span of the matrix columns
- That's when you know oh-oh invertible!
- If always you uniquely solve  $A\vec{x}$  is  $\vec{b}$
- Or if your columns have no linear dependency
- Or if matrix reduces to identity

Not zero - no no

That makes it not invertible!

Shout out if one of these facts is so...

but you'll need  $n$  square columns and rows:

- Like when your matrix determinant's non-zero

Is when you know oh-oh—that makes it invertible!

**Subspace:** it's linear (linear combinations)

**Subspace basis:** span (full row pivots) + l.i. (full column pivots)

$$\text{reduced } A = \begin{bmatrix} 1 & 2 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

**column space of  $A$** =span of columns  $[c_1 \vec{v}_1 + \dots]$  or equations like  $b_1 - 2b_2 + b_3 = 0$  from  $[A|\vec{b}]$  reduction.

A basis is {column 1, ... of original matrix} - they span the column space via linear combinations and are l.i.

**null space of  $A$** =solutions of  $A\vec{x} = \vec{0}$ . Aug: 
$$\begin{bmatrix} 1 & 2 & 1 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

Parametrize:  $s \begin{bmatrix} -2 \\ 1 \\ 0 \end{bmatrix} + t \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$ . A basis is  $\left\{ \begin{bmatrix} -2 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} \right\}$



- $A\vec{x} = \lambda\vec{x}$ : matrix multiplication to scalar multiplication
- determinant  $(A - \lambda I) = 0$  or Maple for  $\lambda$
- plug in each  $\lambda$  for nullspace of  $(A - \lambda I)$  or use Maple for basis for eigenspace
- $A\vec{x} = \lambda\vec{x}$ :  $A$  keeps eigenvectors  $\vec{x}$  on the same line scaled by  $\lambda$ , so can reason geometrically for transformations

$A := \text{Matrix}([[21/40, 3/20], [19/240, 39/40]]);$

$\text{Eigenvectors}(A); \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}, \begin{bmatrix} -6 & 6 \\ 1 & 19 \end{bmatrix}$

The eigenvector decomposition and trajectory:

$$\begin{bmatrix} \text{Owls}_k \\ \text{Rats}_k \end{bmatrix} = c_1 \left(\frac{1}{2}\right)^k \begin{bmatrix} -6 \\ 1 \end{bmatrix} + c_2 \begin{bmatrix} 6 \\ 19 \end{bmatrix}$$

