

## Last time

- Calculating  $\det(A)$  by row reduction
- Algebraic properties of determinants

Last time, we saw that, for  $A$   $3 \times 3$ ,

$$\det(A) = a_{11} \det \begin{bmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{bmatrix} - a_{12} \det \begin{bmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{bmatrix} \\ + a_{13} \det \begin{bmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{bmatrix},$$

a recursive formula for the  $3 \times 3$  determinant  
in terms of the  $2 \times 2$  determinant.

Laplace expansion Let  $A_{ij}$  be the  $(n-1) \times (n-1)$  matrix obtained by omitting the  $i^{\text{th}}$  row and  $j^{\text{th}}$  column from  $A$ .

(1) Expansion along row  $i$ :

$$\det(A) = \sum_{j=1}^n (-1)^{i+j} a_{ij} \det(A_{ij})$$

(2) Expansion down column  $j$ :

$$\det(A) = \sum_{i=1}^n (-1)^{i+j} a_{ij} \det(A_{ij})$$

Ex

$$A = \begin{bmatrix} 1 & 0 & 1 & 2 \\ 9 & 1 & 3 & 0 \\ 9 & 2 & 2 & 0 \\ 5 & 0 & 0 & 3 \end{bmatrix}$$

$$\left( \begin{bmatrix} + & - & + & - \\ - & + & - & + \\ + & - & + & - \\ \vdots & \vdots & \vdots & \vdots \end{bmatrix} \right)$$

Expand down the 2<sup>nd</sup> column:

$$\begin{aligned} \det(A) &= -0 \cdot \det \begin{bmatrix} 9 & 3 & 0 \\ 9 & 2 & 0 \\ 5 & 0 & 3 \end{bmatrix} + 1 \cdot \det \begin{bmatrix} 1 & 1 & 2 \\ 9 & 2 & 0 \\ 5 & 0 & 3 \end{bmatrix} \\ &\quad - 2 \cdot \det \begin{bmatrix} 1 & 1 & 2 \\ 9 & 3 & 0 \\ 5 & 0 & 3 \end{bmatrix} + 0 \cdot \det \begin{bmatrix} 1 & 1 & 2 \\ 9 & 3 & 0 \\ 9 & 2 & 0 \end{bmatrix} \\ &= 5 \det \begin{bmatrix} 1 & 2 \\ 2 & 0 \end{bmatrix} + 3 \begin{bmatrix} 1 \\ 9 & 2 \end{bmatrix} \\ &\quad - 2 \left( 2 \det \begin{bmatrix} 9 & 3 \\ 5 & 0 \end{bmatrix} + 3 \begin{bmatrix} 1 & 1 \\ 9 & 3 \end{bmatrix} \right) \end{aligned}$$

$$= 5(-4) + 3(-7) - 4(-15) - 6(-6)$$

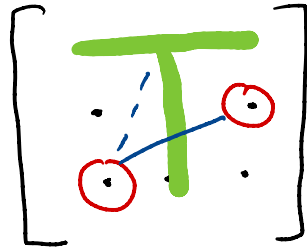
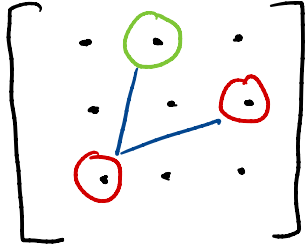
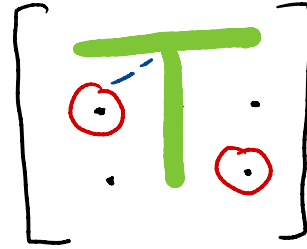
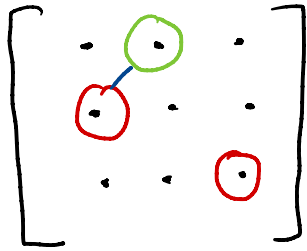
$$= -20 - 21 + 60 + 36$$

$$= 55$$

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Why does this work? For every entry in the  $i^{\text{th}}$  row, collect all the patterns involving that entry, and factor out that entry to get the determinant of the smaller matrix, up to a sign.

accounting for the lost inversions.



Diagonal matrices are good because

- (1) they're easy to work with
- (2) their geometric meaning is transparent
- (3) the systems of equations they represent are "uncoupled", so easy to solve

Matrices that are similar to diagonal matrices are just as good.

Def An  $n \times n$  matrix  $A$  is diagonalizable if  $A$  is similar to a diagonal matrix.

In other words,  $S^{-1}AS$  is diagonal for some invertible  $n \times n$  matrix  $S$ , which is to say there is a basis  $\mathcal{B}$  (the columns of  $S$ ) for  $\mathbb{R}^n$  such that the  $\mathcal{B}$ -matrix of  $A$  is diagonal.

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What does this mean, exactly?

Suppose

$$S = \begin{bmatrix} | & & | \\ \vec{v}_1 & \dots & \vec{v}_n \\ | & & | \end{bmatrix} \quad S^{-1}AS = \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix} = D$$

In this case,

$$[A\vec{v}_1 \dots A\vec{v}_n] = AS = SD = [\lambda_1\vec{v}_1 \dots \lambda_n\vec{v}_n].$$

Def An eigenvector of  $A$  is a vector  $\vec{v} \neq \vec{0}$   
such that

$$A\vec{v} = \lambda\vec{v}$$



for some scalar  $\lambda$ , called the eigenvalue associated to  $\vec{v}$ . A basis  $\{\vec{v}_1, \dots, \vec{v}_n\}$  for  $\mathbb{R}^n$  is called an eigenbasis for  $A$  if each  $\vec{v}_i$  is an eigenvector of  $A$ .

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diagonalizable  $\iff$  an eigenbasis exists

More specifically, if  $\{\vec{v}_1, \dots, \vec{v}_n\}$  are linearly independent eigenvectors of  $A$

with eigenvalues  $\lambda_1, \dots, \lambda_n$ , then

$$S^{-1}AS = \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix} \quad S = \begin{bmatrix} \vec{v}_1 & & \\ & \ddots & \\ & & \vec{v}_n \end{bmatrix}$$

Ex Every nonzero vector is an eigenvector for  $I_n$  with eigenvalue 1.

Ex Any nonzero vector  $\vec{m}$  in  $V$  is an eigenvector for  $\text{proj}_V$  with eigenvalue 1.  
Any nonzero vector  $\vec{m}$  in  $V^\perp$  is an

eigenvector with eigenvalue 0.

Ex An eigenvector with eigenvalue 0 is the same thing as a nonzero vector in  $\ker(A)$ .