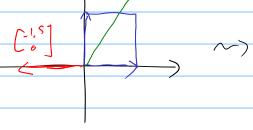
Eigenvalues + eigenvectors.

$$ex$$
 $A = \begin{bmatrix} 3 & 0 \\ 0 & -1 \end{bmatrix}$, Then $A \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 3 \\ 0 \end{bmatrix} = 3 \begin{bmatrix} 1 \\ 0 \end{bmatrix}$.



The x-axis & y-axis consist of eigenvectors.

In fact, these are the only eigenvectors.

ex.
$$B = \begin{bmatrix} \frac{1}{3} & \frac{1}{3} \\ \frac{1}{3} & \frac{1}{5} \\ \frac{1}{3} & \frac{1}{3} \\ \frac{1}{3} & \frac{1}{5} \\ \frac{1}{3} & \frac{1}{3} \\ \frac{1}{3} & \frac{1}{5} \\ \frac{1}{3} & \frac{1}{3} \\ \frac{1}{3} & \frac{1}$$

ex.
$$C = \begin{bmatrix} 2 & 0 \end{bmatrix}$$
 Fvery vector X is an eigenvector C_1 with eigenvalue 2.

ex. X is an eigenvector with eigenvalue O for some A C) AX= 0.X = 0 A has at least one eigenvector with eigenvalue O

A has a nontrivial null space

A is not invertible.

From The set of eigenvectors of A with eigenvalve) is a subspace of R. Proof. Suppose X, Y have eigenvalue). Then $A \cdot (X+Y) = A \cdot X + A \cdot Y = \lambda \cdot X + \lambda \cdot Y = \lambda \cdot (X+Y)$ Suppose X has eigenvalue & and CER Then $A \cdot (cX) = c \cdot AX = c \cdot \lambda X = \lambda \cdot (cX)$ DEigenvectors with distinct eigenvalues generally don't form a subspace. If X has eigenvalue λ , Y has eigenvalue μ , for some A, then $A - (X + Y) = AX + AY = \lambda X + \mu Y$.

This is probably not a scalar multiple of X + Y. Prop. Eigenvectors with different eigenvalues are linearly independent. Proof. Say X has eigenvalue X, Y has eigenvalue M. If X and Y are linearly dependent, then

let's say X = c Y But then X would also have

eigenvalue M. $\rightarrow \leftarrow$ 50 X, Y are linearly independent. Cor. At most n possible eigenvalues for an nxn matrix.

Proof Any eigenvalue has at least one eigenvector ... so has at least a 1-dim'l space of eigenvectors The eigenvectors for distinct eigenvalues are LI from each other, so In of these spaces of circumvectors.

Finding eigenvalues

Suppose
$$\lambda$$
 is an eigenvalue of A . (with X an eigenvector) $AX = \lambda X = (\lambda I) X$.

$$\begin{array}{c} AX - (\lambda I)X = 0 \\ (A - \lambda I)X = 0 \end{array}$$

i.e.
$$det(A - \lambda I) = 0$$
.

$$P_A(\lambda) = det(A - \lambda I)$$

$$P_A(\lambda) = \det(A - \lambda I)$$
.
If A is nxn, this is a degree n polynomial.

$$P_{A}(\lambda) = \begin{bmatrix} y_3 & 4/3 \\ 8/3 & 5/3 \end{bmatrix} - \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix} = \begin{bmatrix} 1/3 - \lambda & 4/3 \\ 8/3 & 5/3 - \lambda \end{bmatrix}$$

$$= (\frac{1}{3} - \lambda)(\frac{5}{3} - \lambda) - (\frac{4}{3}) \cdot (\frac{8}{3})$$

$$= \lambda^{2} - 2\lambda - 3. = (\lambda - 3)(\lambda + 1)$$

Zeros are
$$\lambda = 3$$
, $\lambda = -1$ These are the eigenvalues of A .

ex.
$$D = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$$
 $D - \lambda I = \begin{bmatrix} 2 - \lambda & 0 \\ 0 & 2 - \lambda \end{bmatrix}$

$$P_{D}(\lambda) = |2-\lambda|_{D} = (2-\lambda)^{2} = (\lambda-2)^{2}$$

$$P_{D}(x) = |2-x|^{2} = (x-2)^{2}$$

 $O_{A}(x) = |2-x|^{2} = (x-2)^{2}$
 $O_{A}(x) = |2-x|^{2}$
 O_{A

	More generally, we say that the multiplicity of an
	Pier avalve & is V if the characteristic solumni
	eigenvalue to is v if the characteristic polynomics divisible by $(1-1)^{r_{2}}$ (and not $(1-1)^{r_{2}}$)
	(1) (1) (1) (1)
	What we know about factorizing polynomials
	Fundamental Thm of Algebra.
	Fundamental Thm of Algebra. A degree in polynomial P(1) has exactly in roots.
	if you count with multiplicity
	and you include complex roots.
	In other words,
	In other words, $P(\lambda) = C(\lambda - \lambda_1)^{n_1} (\lambda - \lambda_2)^{n_2} - \cdots (\lambda - \lambda_k)^{n_k}$
	where ,, \k may be complex, and
	$n, + n_2 + \cdots + n_k = n$
	_
	Cor. An nxn matrix has exactly n eigenvalues
	if you count with multiplicity
_	if you count with multiplicity and you include complex eigenvalues.
	A few more useful reminders:
_	A guadratic polynomial ran always be factorized, using the guadratic formula if necessary. Higher-degree polynomials can be factorized numerically by computer but by hand it's pretty hand/impossible if the polynomial has real coefficients, then
	using the gradiatic formula if necessary
_	-> Higher-degree polynomials can be factorized numerically
	by computer, but by hand it's pretty hand/impossible
_	if the polynomial has real coefficients, then
	any non-real roots occur in conjugate pairs. (if on+bi is a root w/ multiplicity k then a-b; is also a root w/ multiplicity k.)
	(it atbi is a root w/ multiplicity k
	then a-b; is also a root w/ multiplicity k.)

ex.
$$U = \begin{bmatrix} 2 & 1 & -1 \\ 0 & 2 & 5 \\ 0 & 0 & 3 \end{bmatrix}$$
. $\begin{vmatrix} 2-\lambda & 1 & -1 \\ 0 & 2-\lambda & 5 \\ 0 & 0 & 3-\lambda \end{vmatrix}$

$$= (2-\lambda)(2-\lambda)(3-\lambda).$$
So the eigenvalues of U are U (double) U 3.

(onclusion: eigenvalues of an upper triangular matrix are the entries on the diagonal.

(Same for lower triangular matrices.)

ex. $R = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$.

PR(X) = $\begin{bmatrix} -\lambda & -1 \\ 1 & -\lambda \end{bmatrix}$ = $\begin{bmatrix} \lambda^2 + 1 \\ 1 & -\lambda \end{bmatrix}$
So $\begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 & -1 \\ 1 \end{bmatrix} = \begin{bmatrix} -1 \\ 1 \end{bmatrix} = \begin{bmatrix} -1 \\ 1 \end{bmatrix} = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$.

So, $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$ is an eigenvector of R with eigenvalue R . R if R is an eigenvector of R with eigenvalue R . R is an eigenvector of R with eigenvalue R . R is an eigenvector of R with eigenvalue R . R is an eigenvector of R with eigenvalue R only eigenvectors are R is R of R only eigenvectors are R is R of R only eigenvectors are R is R of R in R only eigenvectors are R of R is R only eigenvectors are R of R in R of R in R only eigenvectors are R of R in R of R in R of R in R of R only eigenvectors are R of R of R in R of R in R of R only eigenvectors are R of R or R or R or R or R of R or R

Finding eigenvectors.

Recall: The eigenvalues of a matrix A are

the roots of the characteristic polynomial

PA(L) = det(A - LI).

PA(L) = det (A - LI).

If A is nxn, then are n of these, counting with multiplicity & including complex eigenvalves.

Suppose X is an eigenvector of A with eigenvalue X. Then

 $AX = \lambda X = (\lambda I)X$

 $(A - \lambda I) X = 0.$ This is just a system of equations that we can solve for X.

$$P_{A}(\lambda) = \begin{vmatrix} -2 - \lambda & 0 & 1 \\ -2 & -1 & 2 \end{vmatrix}$$

$$P_{A}(\lambda) = \begin{vmatrix} -2 - 1 - \lambda & 2 \\ 2 & -1 & 0 \end{vmatrix}$$

$$= (-2-\lambda) | -1-\lambda | 2 | + 1 | -2 | -1-\lambda |$$

$$= (-2-\lambda) | (-1-\lambda)(-\lambda) + 2 | + | 2 | -2 | (-1-\lambda) |$$

$$= (-2-\lambda) | (-1-\lambda)(-\lambda) + 2 | + | 2 | -2 | (-1-\lambda) |$$

$$= (-2-\lambda) | (\lambda^2 + \lambda + 2) | + | (2\lambda + 4) |$$

$$= -\lambda | (\lambda^2 + 3\lambda + 2) |$$

$$= -\lambda | (\lambda + 2) | (\lambda + 1) |$$

The eigenvalues: $\lambda = 0, -2,$ and -1.

ex [2 0]
$$A = (\lambda - 2)$$
.

Cigenvalves: $\lambda = 2$ (double eigenvalve).

(A-2I) $X = 0$
 $A - 2I = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix} - \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$

So any X is a solution.

ex. $\begin{bmatrix} 1 & 1 \\ 5 = 0 & 1 \end{bmatrix}$ eigenvalves: $\lambda = 1$ (double eigenvalve)

(S-I) $X = 0$
 $\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} X = 0$. $\Rightarrow X = c \begin{bmatrix} 1 \\ 0 \end{bmatrix}$

In this case 1 is a double eigenvalve but has only a 1-dim'l space of eigenvectors.

In general: a k-fold eigenvalve can have a space of eigenvectors of dimension anywhere between 1 and k.

Exercise: Find 3x3 matrices with triple eigenvalves and (a) a 1-dim'l space of eigenvectors, (b) a 2-dim'l space of eigenvectors, (b) a 2-dim'l space of eigenvectors, (b) a 2-dim'l space of eigenvectors, (c) a 3-dim'l space of eigenvectors, (d) a 3-dim'l space of eigenvectors, (e) a

ex.
$$R = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$$
 $P_{R}(\lambda) = \lambda^{2} + 1$

$$X = iy =$$
 $\begin{bmatrix} X \\ y \end{bmatrix} = C \begin{bmatrix} i \\ 1 \end{bmatrix}, C \in C.$

Diagonalization.

Say that A is an nxn matrix representing $T_A: \mathbb{R}^n \to \mathbb{R}^n$ Suppose that there's a basis $B = \{X, --, X, \}$ for \mathbb{R}^n such that each X_i is an eigenvector for A_i with eigenvalue X_i .

What's the matrix for T_A in terms of the basis B_i^2 .

The transformation of T_A in terms of T_A in T_A in

$$A = P_B \cdot D \cdot (B)$$

$$A = P_B \cdot D \cdot P_B \quad \text{where} \quad P_B = \begin{bmatrix} 1 & 1 \\ X, \dots, X \end{bmatrix}$$

Def, A is diagonalizable if then's a basis for R^ consisting of eigenvectors for A.

$$\underbrace{A} = \begin{bmatrix} -2 & 0 & 1 \\ -2 & -1 & 2 \\ 2 & -1 & 0 \end{bmatrix}$$

$$\lambda = -2$$

$$\lambda = -1$$

$$\lambda = 0$$

$$P_{B} = \begin{bmatrix} 1 & 1 & 1 \\ 2 & 3 & 2 \\ 0 & 1 & 2 \end{bmatrix}$$

 $\begin{bmatrix}
1 & 0 & 1 & 3 & -1 & 0 \\
0 & 1 & 0 & 1 -2 & 1 & 0
\end{bmatrix}$ $\begin{bmatrix}
1 & 0 & 1 & 3 & -1 & 0 \\
0 & 1 & 0 & 1 -2 & 1 & 0
\end{bmatrix}$ $\begin{bmatrix}
0 & 0 & 2 & 2 & -1 & 1
\end{bmatrix}$ $\begin{bmatrix}
1 & 0 & 0 & 2 & -1/2 & -1/2 \\
0 & 1 & 0 & -2 & 1 & 0
\end{bmatrix}$ $\begin{bmatrix}
2 & -1/2 & -1/2 & 7 \\
2 & -1/2 & -1/2 & 7
\end{bmatrix}$ $\begin{bmatrix}
0 & 1 & 0 & 1 & -1/2 & 1/2
\end{bmatrix}$ $\begin{bmatrix}
0 & 0 & 1 & 1 & -1/2 & 1/2
\end{bmatrix}$ $\begin{bmatrix}
1 & 0 & 0 & 2 & -1/2 & -1/2
\end{bmatrix}$ $\begin{bmatrix}
1 & 0 & 0 & 2 & -1/2 & -1/2
\end{bmatrix}$ $\begin{bmatrix}
1 & 0 & 0 & 2 & -1/2 & -1/2
\end{bmatrix}$ $\begin{bmatrix}
2 & -1/2 & -1/2
\end{bmatrix}$ $\begin{bmatrix}
2 & -1/2 & -1/2
\end{bmatrix}$ $\begin{bmatrix}
1 & -1/2 & 1/2
\end{bmatrix}$ $\begin{bmatrix}
1 & -1/2 & 1/2
\end{bmatrix}$ - R2 + R1 Application: A = (PBDP2) 500

= PBDP21 . PBDP21

= PBD 500 PE1

= PBD 500 PE1 PB (-2) 500 0 0 0 When is a matrix diagonalizable? Prop. If A has a distinct eigenvalues, then it's diagonalizable

Proof Each eigenvalue has at least one eigenvectory
and these are LI for distinct eigenvalues.

So there's a basis for IR consisting of
one eigenvector for each eigenvalue.

$$\begin{array}{lll} \text{Eigenvalues: } \lambda = 1 & \text{double} \end{array}) \\ \text{Eigenvalues: } \lambda = 1 & \text{double} \end{array}) \\ \text{Eigenvalues: } \lambda = \left[\begin{array}{c} 1 \\ 0 \end{array}\right]. \\ \text{Eigenvalues: } \lambda = \left[\begin{array}{c} 1 \\ 0 \end{array}\right]. \\ \text{There's no basis of } R^2 & \text{consisting of eigenvalues}. \\ \text{Solidions Solidions} \end{array}$$

$$\begin{array}{lll} \text{Expansion of } R^2 & \text{consisting of eigenvalues}. \\ \text{Expansion of } R & \text{consi$$

$$B = \begin{cases} 2 & 2 & 3 \\ \frac{3}{2} + \frac{133}{2} & \frac{3}{2} - \frac{133}{2} \end{cases}$$

$$P_{B} = \begin{cases} 2 & 2 & 4 + (P_{B}) = 2(\frac{3}{2} - \frac{13}{2}) \\ \frac{3}{2} + \frac{133}{2} & \frac{3}{2} - \frac{133}{2} & -2(\frac{3}{2} + \frac{133}{2}) \\ -2(\frac{3}{2} + \frac{133}{2}) & \frac{3}{2} - \frac{133}{2} & -2 \end{cases}$$

$$C_{B} = P_{B}^{-1} = \frac{1}{-2\sqrt{3}3} \begin{bmatrix} \frac{3}{2} + \frac{133}{2} & -2 \\ -\frac{3}{2} - \frac{133}{2} & 2 \end{bmatrix}$$

$$C_{B} = P_{B}^{-1} = \frac{1}{-2\sqrt{3}3} \begin{bmatrix} \frac{3}{2} + \frac{133}{2} & 2 \\ -\frac{3}{2} - \frac{133}{2} & 2 \end{bmatrix}$$

$$C_{B} = P_{B}^{-1} = \frac{1}{-2\sqrt{3}3} \begin{bmatrix} \frac{3}{2} + \frac{133}{2} & 2 \\ -\frac{3}{2} - \frac{133}{2} & 2 \end{bmatrix}$$