

**Dynamical Systems:** If  $A$  is an  $n \times n$  matrix, and  $\vec{x}_0 \in \mathbb{R}^n$  we can create a *dynamical system* of the form

$$\vec{x}_0, \vec{x}_1, \vec{x}_2, \vec{x}_3, \dots \quad \text{where} \quad \vec{x}_{t+1} = A\vec{x}_t \quad (\text{recursive definition})$$

Many applications are done this way, with  $t$  typically representing time: population dynamics, weather models, financial models, historical processes, pharmacology, ....

Note: we get to  $\vec{x}_k$  by applying  $A$  to  $\vec{x}_0$   $k$  times. Thus:  $\vec{x}_k = A^k x_0$ .

1. **Today's CheckPoint.** A dynamical system is defined by the recursive rule  $x_{k+1} = A\vec{x}_k$  with matrix  $A$  and initial value  $\vec{x}_0$ :

$$A = \begin{bmatrix} 97/100 & 3/55 \\ -4/55 & 123/100 \end{bmatrix}, \quad \vec{x}_0 = \begin{bmatrix} 1 \\ 15 \end{bmatrix}$$

Can generate some values

$$\begin{bmatrix} 1 \\ 15 \end{bmatrix}, \begin{bmatrix} 1.7 \\ 16.7 \end{bmatrix}, \begin{bmatrix} 2.41 \\ 18.55 \end{bmatrix}, \begin{bmatrix} 3.137 \\ 20.567 \end{bmatrix}, \begin{bmatrix} \phantom{0} \\ \phantom{0} \end{bmatrix}, \dots \begin{bmatrix} \phantom{0} \\ \phantom{0} \end{bmatrix}, \dots$$

$\vec{x}_0 \quad \vec{x}_1 \quad \vec{x}_2 \quad \vec{x}_3 \quad \vec{x}_4 \quad \vec{x}_k$

A “closed formula” is a formula for the  $k$ th term that can be computed using  $k$  and doesn't require us to compute all of the values up to  $\vec{x}_k$  first.

The *eigenvectors* and *eigenvalues* of  $A$  are what we need:

$$\lambda_1 = 1.1 \quad \lambda_2 = 0.9$$

$$\vec{v}_1 = \begin{bmatrix} 1 \\ 4 \end{bmatrix} \quad \vec{v}_2 = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$$

step 1: Write  $\vec{x}_0$  as a linear combination of the eigenbasis (how?): ① augment and row reduce  
② use the inverse of  $\begin{bmatrix} 1 & 3 \\ 4 & 1 \end{bmatrix}$   
③ Guess and check

$$\begin{bmatrix} 1 \\ 15 \end{bmatrix} = \underline{4} \begin{bmatrix} 1 \\ 4 \end{bmatrix} + \underline{-1} \begin{bmatrix} 3 \\ 1 \end{bmatrix}$$

step 2: Apply  $A$   $k$  times:

$$\vec{x}_k = \begin{bmatrix} a_k \\ b_k \end{bmatrix} = A^k \vec{x}_0 = A^k \begin{bmatrix} 1 \\ 15 \end{bmatrix} = \underline{4} A^k \begin{bmatrix} 1 \\ 4 \end{bmatrix} + \underline{(-1)} A^k \begin{bmatrix} 3 \\ 1 \end{bmatrix} = \underline{4} (1.1)^k \begin{bmatrix} 1 \\ 4 \end{bmatrix} + \underline{-1} (0.9)^k \begin{bmatrix} 3 \\ 1 \end{bmatrix}$$

Thus,

$$a_k = 4 (1.1)^k - 3 (0.9)^k$$

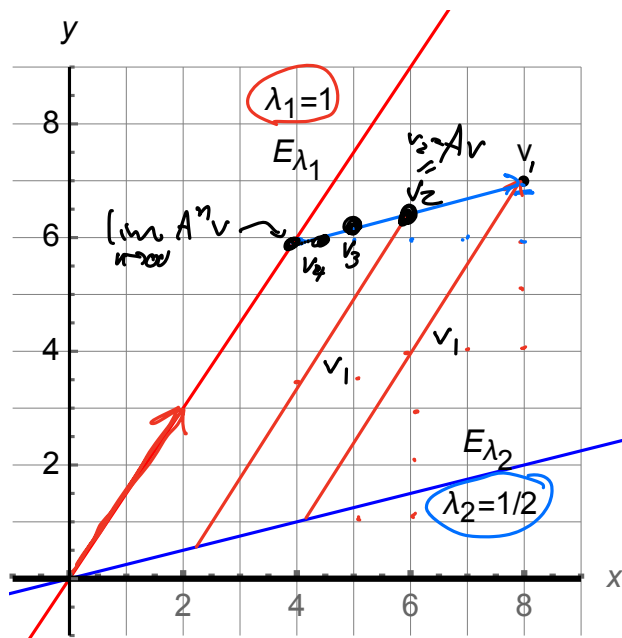
$$b_k = 16 (1.1)^k - (0.9)^k$$

Alternatively: diagonalize:  $A = PDP^{-1} = \begin{bmatrix} 1 & 3 \\ 4 & 1 \end{bmatrix} \begin{bmatrix} 1.1 & 0 \\ 0 & 0.9 \end{bmatrix} \begin{bmatrix} 1 & -3 \\ -4 & 1 \end{bmatrix}$

$$A^k \begin{bmatrix} 1 \\ 15 \end{bmatrix} = \begin{bmatrix} 1 & 3 \\ 4 & 1 \end{bmatrix} \begin{bmatrix} 1.1^k & 0 \\ 0 & 0.9^k \end{bmatrix} \begin{bmatrix} 1 & -3 \\ -4 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 15 \end{bmatrix} = \begin{bmatrix} 1 & 3 \\ 4 & 1 \end{bmatrix} \begin{bmatrix} 1.1^k & 0 \\ 0 & 0.9^k \end{bmatrix} \begin{bmatrix} 4 \\ -1 \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 3 \\ 4 & 1 \end{bmatrix} \begin{bmatrix} 4(1.1)^k \\ -(0.9)^k \end{bmatrix} = 4(1.1)^k \begin{bmatrix} 1 \\ 4 \end{bmatrix} - (0.9)^k \begin{bmatrix} 3 \\ 1 \end{bmatrix} \text{ as above}$$

2.  $A$  is a matrix with eigenvalues  $\lambda_1 = 1$  and  $\lambda_2 = 1/2$  and corresponding eigenspaces shown below. The vector  $\vec{v}$  is shown on the plot at position  $(8, 7)$ . Using the eigeninformation, compute the position of  $A\vec{v}$  and plot it on the graph. Compute it exactly if possible; if not, estimate it.

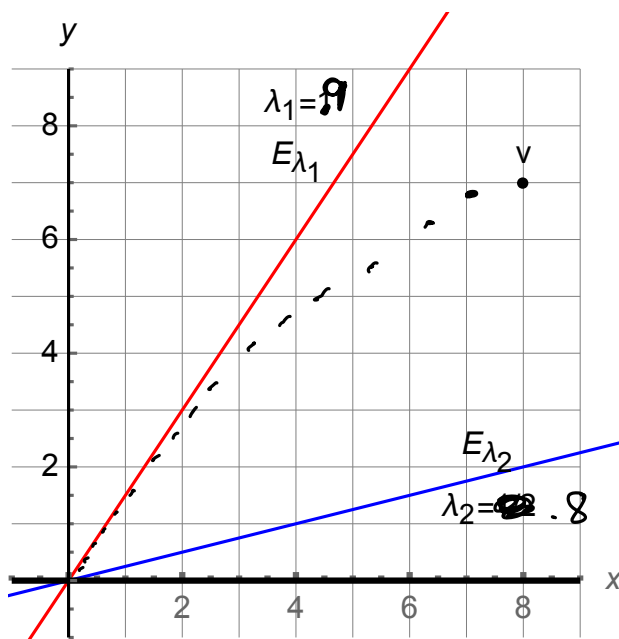
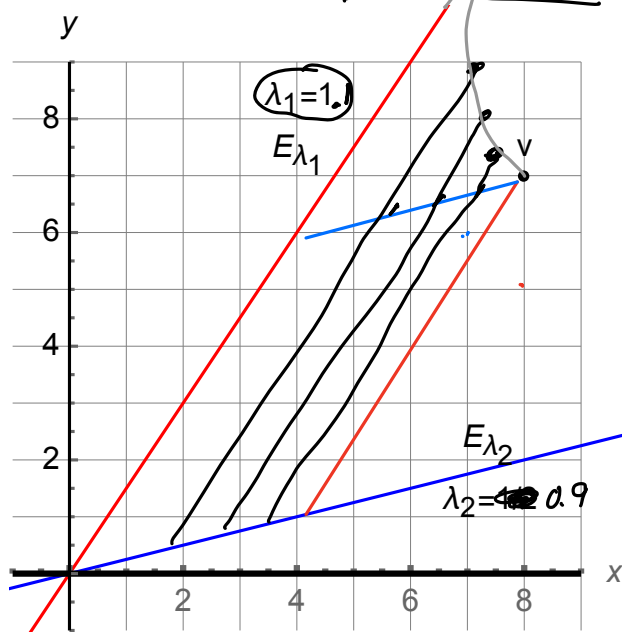


$A\vec{v}$

$$\begin{aligned} \vec{v} &= \vec{v}_1 + \vec{v}_2 \\ A\vec{v} &= A\vec{v}_1 + A\vec{v}_2 \\ &= \vec{v}_1 + \frac{1}{2}\vec{v}_2 \end{aligned}$$

On this same problem, where will  $A^2\vec{v}$  be?

3. What if  $\lambda_1$  is changed to  $\lambda_1 = 1.1$  or  $\lambda_1 = 0.9$ ?



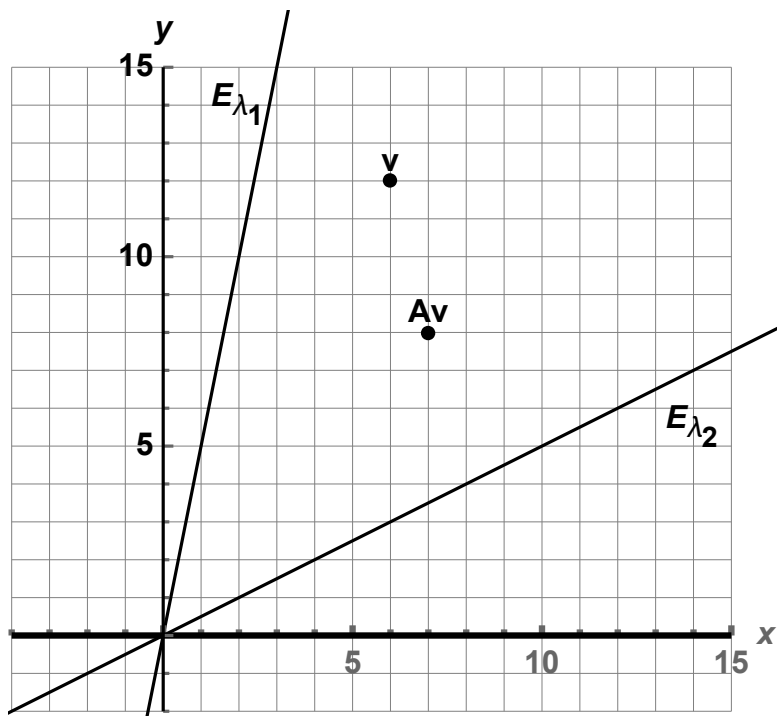
4.  $A$  is a matrix that sends  $\vec{v}$  to  $A\vec{v}$  as shown in the plot below with its two eigenspaces  $E_{\lambda_1}$  and  $E_{\lambda_2}$ .

(a) Estimate, as accurately as possible from the given information, the eigenvalues  $\lambda_1 = \underline{\hspace{1cm}}$  and  $\lambda_2 = \underline{\hspace{1cm}}$ .

(b) Indicate on the plot above where  $A^2\vec{v}$  will be.

(c) What happens in the limit:  $\lim_{n \rightarrow \infty} A^n \vec{v}$ ?

(d) If  $A^n \vec{v} = \begin{bmatrix} x_n \\ y_n \end{bmatrix}$  what happens to the ratio  $x_n/y_n$  as  $n$  grows larger and larger?



5. **Dominant Eigenvectors** It is very common for an  $n \times n$  matrix  $A$  to have an eigenvalue  $\lambda_1$  that is bigger, in absolute value, than all of the other eigenvalues.

$$\begin{aligned}
 &A \quad |\lambda_1| > |\lambda_2| \geq |\lambda_3| \geq \dots \geq |\lambda_n| \\
 &X = c_1 v_1 + c_2 v_2 + c_3 v_3 + \dots + c_n v_n \\
 &A^n X = c_1 \lambda_1^n v_1 + c_2 \lambda_2^n v_2 + c_3 \lambda_3^n v_3 + \dots + c_n \lambda_n^n v_n \\
 &= \lambda_1^n \left[ c_1 v_1 + c_2 \left( \frac{\lambda_2}{\lambda_1} \right)^n v_2 + c_3 \left( \frac{\lambda_3}{\lambda_1} \right)^n v_3 + \dots + c_n \left( \frac{\lambda_n}{\lambda_1} \right)^n v_n \right] \\
 &\quad \quad \quad \underbrace{\hspace{10em}}_{\rightarrow 0} \\
 &\approx \lambda_1^n c_1 v_1
 \end{aligned}$$

converges to the direction of the dominant eigenvector

6. Look at the Northern Spotted Owl example on R: