

Eigen problems:

Calculating Eigen

— we know ~~how~~ what eigenvectors are and how to calculate it.

— we can combine this idea with concept of changing basis (Covered earlier)

→ what emerges is powerful tool. for performing <sup>efficient</sup> matrix operations, called diagonalization.

Sometimes we need to apply the same matrix operator MANY TIMES

Eg Transformation matrix  $T$ , that represents the change in location of particle after a single time stamp

$$T = \begin{bmatrix} 0.9 & 0.8 \\ -1 & 0.35 \end{bmatrix}$$

$$V_0 = \begin{bmatrix} 0.5 \\ 1 \end{bmatrix}$$



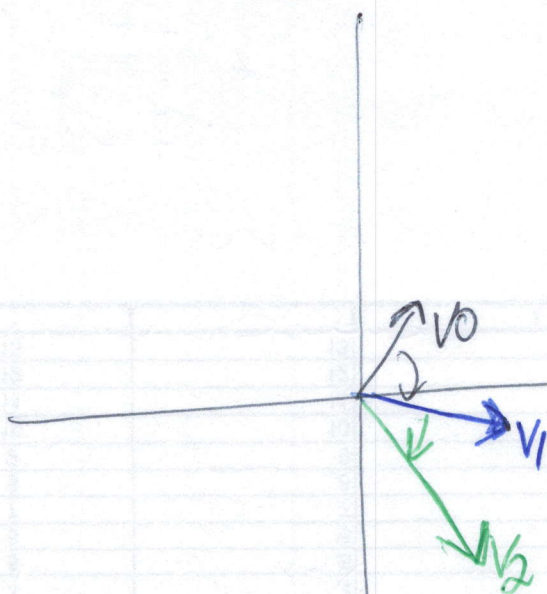
Initial position  $V_0$ , multiplied by transformation  $T$   
 gives us a new location

②

$$V_1 = TV_0$$

$$V_2 = TV_1$$

↳ same as:  
 $V_2 = T(TV_0)$



to work out where our particle will be after 2 timesteps  
 we can find  $V_2$  by simply multiplying  $V_1$  by  $T$ .  
 which is the same thing as multiplying

$V_0$  by  $T$  2 times

$$V_2 = TV_1$$

$$\therefore V_2 = T(TV_0)$$

$$\therefore V_2 = T^2 V_0$$

( $T$  squared times  $V_0$ )

Imagine we expect the same linear  
 transformation operation to occur everytime  
 for  $n$  timesteps.



$$\therefore V_2 = T^2 V_0$$

$$\therefore V_n = T^n V_0$$

- We've seen the amount of work it takes to apply a single transformation<sup>3D</sup> matrix multiplication.

So if we were to imagine that  $T$  tells us what happens in 1 second, and we want to know where our particle is in weeks from now, then  $n$  will be 1.2 million

i.e. we will need to multiply  $T$  by itself 1 million times, which will

take a while. -----

But if all the terms (~~are zero~~) in matrix is zero, except those in ~~the~~ leading diagonal, we refer to it as a diagonal matrix.  $\rightarrow$  and when raising matrices to powers, diagonal matrices make things a lot easier.



(4)

∴ all you need to do is put all terms in the diagonal to power of  $n$ , and then we'll get the answer.

$$\therefore T^n = \begin{bmatrix} a^n & 0 & 0 \\ 0 & b^n & 0 \\ 0 & 0 & c^n \end{bmatrix} \Rightarrow \text{simple} \dots$$

But what if  $T$  is not a diagonal matrix.  
(answer will come from Eigen analysis)

To solve this, we'll change to a basis, where our transformation  $T$ , becomes diagonal.  
 $\Rightarrow$  which is called an Eigenbasis.

— we can then easily apply our power of  $n$  to the diagonalize form, and finally transforming the resultant matrix back again.

Quing →  $T$  to power of  $n$ , but avoiding much of work.

Remember when changing Basis, each column of a transformed matrix is



represents the new location of the transformed ⑤  
unit vectors

To build an Eigen<sup>Basis</sup> Conversion matrix,  
we just plug in each of our eigenvectors as columns

$C = \text{Eigenvector } 1, 2, 3$ , but using a 3D example

$$\therefore C = \begin{bmatrix} x_1 & x_2 & x_3 \\ \vdots & \vdots & \vdots \\ 1 & 1 & 1 \end{bmatrix} \quad D = \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{bmatrix}$$

(So now multiplying by  $T$  is just a pure scalar  
multiplication, it can now be represented  
by a diagonal matrix.

$\Rightarrow$  Crucially the diagonal matrix  $D$  contains  
the corresponding Eigenvalues of matrix  $T$ .

So close to unleashing power of EIGEN ... Final  
link to see the following:

① Applying the transformation  $T$ , is just the same  
as converting to our Eigen Basis,

② applying the diagonalizing matrix

③ and then converting back again



(6)

$$\therefore T = C \Delta C^{-1}$$

$$T^2 = C \cancel{\Delta C^{-1}} \cancel{C \Delta C^{-1}}$$

[But multiplying a matrix, then multiplying it with its inverse, is the same as doing nothing at all]  
 $\Rightarrow$  'So we can just remove this operation.

$$= C \Delta \Delta C^{-1}$$

$$= C \Delta^2 C^{-1}$$

We can then generalize this to every power of  $T$ .

$$\therefore T^n = C \Delta^n C^{-1}$$

$$\begin{array}{ccc} V & \xrightarrow{T^n} & T^n V \\ C^{-1} \downarrow & & \uparrow C \\ V_E & \longrightarrow & [T^n]_E \end{array}$$

We have a method that allows us to apply a transfer matrix as many times as we like, without paying a large computational cost.