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Lecture09a

Similarity Transformation and Diagonalization



Lecture is on YouTube

The YouTube video entitled 'Similarity Transformation and Diagonalization' that covers this lecture is located at <https://youtu.be/wvRlvDYDIgw>.

Outline

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Similarity Transformation

The mathematical definition of a similarity transformation within the context of linear algebra (https://en.wikipedia.org/wiki/Matrix_similarity) concerns two $n \times n$ matrices, A and \tilde{A} are called similar if there exists an invertible $n \times n$ matrix T such that

$$\tilde{A} = T^{-1} A T \quad (\text{Eq.1})$$

As the name suggests, the two matrices A and \tilde{A} are similar in the sense that they share some similar (or closely related) properties.

Property 1 : Same determinant

Recall the determinant of a matrix product (using notation that $|M| = \det(M)$) has the property that $|AB| = |A| \cdot |B|$. So applied to our similarity transformation we can write

$$\begin{aligned}
 |\tilde{A}| &= |T^{-1}AT| \\
 &= |T^{-1}| \cdot |A| \cdot |T| \text{ recall: } |M^{-1}| = |M|^{-1} \\
 &= |T|^{-1} \cdot |A| \cdot |T| \\
 &= \frac{1}{|T|} \cdot |A| \cdot |T|
 \end{aligned}$$

$$|\tilde{A}| = |A| \quad (\text{Eq.2})$$

So we see that the determinant of the matrix is not changed by the similarity transformation.

Property 2 : Same characteristic equations (and therefore same eigenvalues)

Recall the definition of eigenvalues/eigenvectors is

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

$$(A - \lambda I)\bar{v} = \bar{0}$$

So eigenvalues are found by solving characteristic equation of

$$|A - \lambda I| = 0$$

Let's rewrite this using the fact that $T\tilde{A}T^{-1} = A$

$$\begin{aligned}
 |A - \lambda I| &= |T\tilde{A}T^{-1} - \lambda I| \\
 &= |T\tilde{A}T^{-1} - \lambda TT^{-1}| \\
 &= |T\tilde{A}T^{-1} - T(\lambda I)T^{-1}| \\
 &= |T(\tilde{A} - \lambda I)T^{-1}| \quad \text{recall: } |AB| = |A| \cdot |B| \\
 &= |T| \cdot |\tilde{A} - \lambda I| \cdot |T^{-1}| \quad \text{recall: } |A^{-1}| = |A|^{-1} \\
 &= |T| \cdot |\tilde{A} - \lambda I| \cdot |T|^{-1}
 \end{aligned}$$

$$= \left| T \right| \cdot \left| \tilde{A} - \lambda I \right| \cdot \frac{1}{\left| T \right|}$$

$$\left| A - \lambda I \right| = \left| \tilde{A} - \lambda I \right|$$

Since the characteristic equation of A and \tilde{A} are the same, eigenvalues don't change under a similarity transformation. This holds true even with multiplicity.

$$\text{eigenvalues}(\tilde{A}) = \text{eigenvalues}(A) \quad (\text{Eq.3})$$

Property 3 : Similar eigenvectors

Suppose that λ_i is an eigenvalue of A and \bar{v}_i is the corresponding vector

λ_i, \bar{v}_i = an eigenvalue and eigenvector pair of A

We can multiply \tilde{A} by an arbitrary vector \bar{u}_i

$$\tilde{A} \bar{u}_i = T^{-1} A T \bar{u}_i \quad \text{let } \bar{u}_i = T^{-1} \bar{v}_i$$

$$= T^{-1} A T T^{-1} \bar{v}_i$$

$$= T^{-1} A \bar{v}_i \quad \text{recall by the definition of an eigenvalue/eigenvector } A \bar{v}_i = \lambda_i \bar{v}_i$$

$$= T^{-1} \lambda_i \bar{v}_i$$

$$\tilde{A} \bar{u}_i = \lambda_i T^{-1} \bar{v}_i \quad \text{recall } \bar{u}_i = T^{-1} \bar{v}_i$$

$$\tilde{A} \bar{u}_i = \lambda_i \bar{u}_i$$

So we see that this is the definition of an eigenvalue/eigenvector for \tilde{A} and therefore \bar{u}_i is an eigenvector of \tilde{A} and λ_i is the corresponding eigenvalue of \tilde{A} . We can then note that we previously used

$$\bar{u}_i = T^{-1} \bar{v}_i$$

So we see that the eigenvectors of \tilde{A} are different from the eigenvectors of A but they are related through

$$\text{eigenvectors}(\tilde{A}) = T^{-1} \text{eigenvectors}(A) \quad (\text{Eq.4})$$

Property 4 : Same trace

Recall that the trace of a matrix is simply the sum of the diagonal elements. Trace is only defined for square matrices.

$$\text{trace}(A) = \text{tr}(A) = \sum_{i=1}^n A_{ii}$$

Recall that if A is the product of two matrices ($A = B_{n \times m} C_{m \times n}$) then we can write

$$\text{tr}(B C) = \text{tr}(C B)$$

$$\text{In[]:= Bmat} = \begin{pmatrix} \text{b11} & \text{b12} \\ \text{b21} & \text{b22} \\ \text{b31} & \text{b32} \end{pmatrix}; \text{Cmat} = \begin{pmatrix} \text{c11} & \text{c12} & \text{c13} \\ \text{c21} & \text{c22} & \text{c23} \end{pmatrix};$$

Amat = Bmat.Cmat;

Amat // MatrixForm

Tr[Bmat.Cmat] == Tr[Cmat.Bmat]

Out[]:=MatrixForm=

$$\begin{pmatrix} \text{b11 c11} + \text{b12 c21} & \text{b11 c12} + \text{b12 c22} & \text{b11 c13} + \text{b12 c23} \\ \text{b21 c11} + \text{b22 c21} & \text{b21 c12} + \text{b22 c22} & \text{b21 c13} + \text{b22 c23} \\ \text{b31 c11} + \text{b32 c21} & \text{b31 c12} + \text{b32 c22} & \text{b31 c13} + \text{b32 c23} \end{pmatrix}$$

Out[]:= True

For our similarity transformation we have

$$\begin{aligned} \text{tr}(\tilde{A}) &= \text{tr}(T^{-1} A T) && \text{let } B = T^{-1}, C = A T \\ &= \text{tr}(B C) && \text{recall: } \text{tr}(B C) = \text{tr}(C B) \\ &= \text{tr}(C B) \\ &= \text{tr}(A T T^{-1}) \end{aligned}$$

$$\text{tr}(\tilde{A}) = \text{tr}(A) \quad (\text{Eq.5})$$

So we see that the trace of the matrix is not changed by the similarity transformation.

Property 5 : Same rank

We now consider the rank of \tilde{A} . To start, we can recall that $\text{rank}(A)$ is the dimension of the vector space spanned by the columns of A . This corresponds to the maximal number of linearly independent columns of A . If we multiply the matrix A by an invertible matrix T , this is equivalent to transforming the vector space spanned by the columns but it does not reduce the number of linearly independent vectors. Therefore

$$\begin{aligned} \text{rank}(T A) &= \text{rank}(A T) = \text{rank}(A) && \text{for } T \text{ invertible} && (\text{Eq.6.a}) \\ \text{rank}(T^{-1} A) &= \text{rank}(A T^{-1}) = \text{rank}(A) && \text{for } T^{-1} \text{ invertible} \end{aligned}$$

An alternative way to prove this is to use the fact that $\text{rank}(A B) \leq \min(\text{rank}(A), \text{rank}(B))$ for any two matrices A and B . If $B = T$ is invertible we can write

$$\text{rank}(A T) \leq \min(\text{rank}(A), \text{rank}(T))$$

We note that $\{T \text{ invertible}\} \iff \{\text{rank}(T) = n\}$ and therefore $\text{rank}(T) = n$ must be greater than or equal to $\text{rank}(A)$ so we have

$$\text{rank}(A T) \leq \text{rank}(A) \quad (\text{Eq.6.b})$$

We can make use of Eq.6.b by replacing the matrices A and T with $A \rightarrow A T$ and $T \rightarrow T^{-1}$

$$\text{rank}((A T) T^{-1}) \leq \text{rank}(A T)$$

$$\text{rank}(A) \leq \text{rank}(A T) \quad (\text{Eq.6.c})$$

So we see that the only way that Eq.6.b and Eq.6.c are true simultaneously is if

$$\text{rank}(A T) = \text{rank}(A) \quad (\text{Eq.6.d})$$

The same can be done to show $\text{rank}(T A) = \text{rank}(A)$ which yields Eq.6.a as previously described.

We can repeat this for a matrix T^{-1} to show that

$$\text{rank}(A T^{-1}) = \text{rank}(T^{-1} A) = \text{rank}(A)$$

So for our similar matrix, we can write

$$\text{rank}(\tilde{A}) = \text{rank}(T^{-1} A T)$$

$$= \text{rank}((T^{-1} A) T) \quad \text{recall: } \text{rank}((T^{-1} A) T) = \text{rank}(T^{-1} A) \text{ pursuant to Eq.6.a}$$

$$= \text{rank}(T^{-1} A) \quad \text{recall: } \text{rank}(T^{-1} A) = \text{rank}(A) \text{ pursuant to Eq.6.a}$$

$$\text{rank}(\tilde{A}) = \text{rank}(A) \quad (\text{Eq.6})$$

Diagonalization

One particularly useful transformation matrix is a matrix of eigenvectors of the A matrix. Recall that an eigenvalue and an eigenvector satisfy the following equation

$$A \bar{v}_i = \lambda_i \bar{v}_i$$

$$A \bar{v}_i - \lambda_i \bar{v}_i = \bar{0}$$

Let's write this out as a series of equations ($i = 1, \dots, n$)

$$\begin{aligned}
 A\bar{v}_1 - \lambda_1 \bar{v}_1 &= \bar{0} \\
 A\bar{v}_2 - \lambda_2 \bar{v}_2 &= \bar{0} \\
 &\vdots \\
 A\bar{v}_n - \lambda_n \bar{v}_n &= \bar{0}
 \end{aligned}$$

Stacking each equation into a separate column yields

$$[A\bar{v}_1 \ A\bar{v}_2 \ \dots \ A\bar{v}_n] - [\lambda_1 \bar{v}_1 \ \lambda_2 \bar{v}_2 \ \dots \ \lambda_n \bar{v}_n] = [\bar{0} \ \bar{0} \ \dots \ \bar{0}]$$

$$A[\bar{v}_1 \ \bar{v}_2 \ \dots \ \bar{v}_n] - [\bar{v}_1 \ \bar{v}_2 \ \dots \ \bar{v}_n] \begin{pmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \dots & \dots & \vdots \\ 0 & \dots & \dots & \lambda_n \end{pmatrix} = [\bar{0} \ \bar{0} \ \dots \ \bar{0}]$$

$$A[\bar{v}_1 \ \bar{v}_2 \ \dots \ \bar{v}_n] = [\bar{v}_1 \ \bar{v}_2 \ \dots \ \bar{v}_n] \begin{pmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \dots & \dots & \vdots \\ 0 & \dots & \dots & \lambda_n \end{pmatrix}$$

let $T = [\bar{v}_1 \ \bar{v}_2 \ \dots \ \bar{v}_n]$

$$AT = T \begin{pmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \dots & \dots & \vdots \\ 0 & \dots & \dots & \lambda_n \end{pmatrix}$$

If T is invertible (AKA if the eigenvectors span the space) then we can write

$$\tilde{A} = T^{-1}AT \quad (\text{Eq. 7})$$

where $\tilde{A} = \begin{pmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \dots & \dots & \vdots \\ 0 & \dots & \dots & \lambda_n \end{pmatrix}$ (diagonal matrix with eigenvalues along the diagonal)

$$T = [\bar{v}_1 \ \bar{v}_2 \ \dots \ \bar{v}_n] \quad (\text{vector of eigenvectors in column format})$$

So we see that using this transformation will yield a diagonal \tilde{A} matrix with the eigenvalues on the diagonal. The similarity transform $T = \text{eigenvectors}(A)$ is known as the modal matrix of A .

This helps us understand the result at https://en.wikipedia.org/wiki/Diagonalizable_matrix which states that:

“An $n \times n$ matrix A over a field F is diagonalizable if and only if the sum of the dimensions of its eigenspaces is equal to n , which is the case if and only if there exists a basis of F^n consistent of eigenvectors of A . If such a basis has been found, one can form the matrix P having these basis vectors as columns, and $P^{-1} A P$ will be a diagonal matrix whose diagonal entries are the eigenvalues of A . The matrix P is known as the modal matrix for A .”

Examples

Example 1: Non-Defective Matrix

Consider the following non-defective matrix (meaning it can be diagonalized).

$$\text{In[]:= } \mathbf{A} = \begin{pmatrix} 3 & 4 & 2 & 1 \\ 0 & 1 & -1 & -1 \\ -1 & -1 & 3 & 0 \\ 1 & 1 & -1 & 2 \end{pmatrix};$$

We can use a transformation matrix of

$$\mathbf{T} = \begin{pmatrix} 1 & \frac{2}{7} & 4 & -3 \\ 2 & 1 & 2 & \frac{3}{4} \\ -2 & \frac{2}{3} & \frac{3}{4} & 2 \\ -\frac{5}{6} & -5 & 8 & 6 \end{pmatrix}$$

We can verify that T is invertible

In[]:= Det[T]

17933

Out[]:=
56

We can calculate \tilde{A}

In[]:= Atilde = Inverse[T].A.T;

Atilde // MatrixForm

Atilde // MatrixForm // N

Out[]//MatrixForm=

$$\begin{pmatrix} \frac{235009}{71732} & \frac{768517}{753186} & -\frac{127303}{35866} & -\frac{332845}{71732} \\ -\frac{728105}{430392} & \frac{1653481}{645588} & -\frac{204022}{53799} & \frac{50267}{430392} \\ \frac{14635}{53799} & \frac{24730}{125531} & \frac{75763}{17933} & \frac{24300}{17933} \\ -\frac{245935}{322794} & \frac{1524821}{3389337} & -\frac{927161}{161397} & -\frac{342869}{322794} \end{pmatrix}$$

Out[]//MatrixForm=

$$\begin{pmatrix} 3.27621 & 1.02035 & -3.54941 & -4.64012 \\ -1.69173 & 2.5612 & -3.7923 & 0.116794 \\ 0.272031 & 0.197003 & 4.22478 & 1.35504 \\ -0.761895 & 0.449888 & -5.7446 & -1.06219 \end{pmatrix}$$

We can verify the aforementioned properties.

Property 1: Same determinant

```
In[ ]:= detA = Det[A]
      detAtilde = Det[Atilde]
      detA == detAtilde
```

```
Out[ ]:= 20
```

```
Out[ ]:= 20
```

```
Out[ ]:= True
```

Property 2: Same characteristic equations (and therefore same eigenvalues)

We can verify that $|A - \lambda I| = |\tilde{A} - \lambda I|$

```
In[ ]:= charEqA = Det[A - λ * IdentityMatrix[4]]
      charEqAtilde = Det[Atilde - λ * IdentityMatrix[4]]
      charEqA == charEqAtilde
```

```
Out[ ]:= 20 - 42 λ + 30 λ^2 - 9 λ^3 + λ^4
```

```
Out[ ]:= 20 - 42 λ + 30 λ^2 - 9 λ^3 + λ^4
```

```
Out[ ]:= True
```

Which means the eigenvalues are the same

```
In[ ]:= eigA = Eigenvalues[A]
      eigAtilde = Eigenvalues[Atilde]
      eigA == eigAtilde
```

```
Out[ ]:= {3 + 1i, 3 - 1i, 2, 1}
```

```
Out[ ]:= {3 + 1i, 3 - 1i, 2, 1}
```

```
Out[ ]:= True
```

Property 3: Similar eigenvectors

We can now verify that $\text{eigenvectors}(\tilde{A}) = T^{-1} \text{eigenvectors}(A)$


```
In[ ]:= (*Mathematica's Eigenvectors returns eigenvectors in row format,
transpose to get into col format*)
eigVecA = Transpose[Eigenvectors[A]];
eigVecAtilde = Transpose[Eigenvectors[Atilde]];
```

```
eigVecA // MatrixForm
eigVecAtilde // MatrixForm // N
```

Out[]//MatrixForm=

$$\begin{pmatrix} i & -i & 2 & -9 \\ 0 & 0 & -1 & 5 \\ -1 & -1 & 1 & -2 \\ 1 & 1 & 0 & 2 \end{pmatrix}$$

Out[]//MatrixForm=

$$\begin{pmatrix} 0.191206 + 1.77394 i & 0.191206 - 1.77394 i & 1.72126 & 1.22901 \\ 0.287534 - 2.35959 i & 0.287534 + 2.35959 i & -0.348722 & -0.0135712 \\ -0.709973 - 0.594141 i & -0.709973 + 0.594141 i & -0.788653 & -0.523043 \\ 1. & 1. & 1. & 1. \end{pmatrix}$$

```
In[ ]:= (*Check that we extracted the eigenvectors correctly*)
```

```
v1 = Transpose[{eigVecA[[All, 1]]}];
v2 = Transpose[{eigVecA[[All, 2]]}];
v3 = Transpose[{eigVecA[[All, 3]]}];
v4 = Transpose[{eigVecA[[All, 4]]}];
```

```
vtilde1 = Transpose[{eigVecAtilde[[All, 1]]}];
vtilde2 = Transpose[{eigVecAtilde[[All, 2]]}];
vtilde3 = Transpose[{eigVecAtilde[[All, 3]]}];
vtilde4 = Transpose[{eigVecAtilde[[All, 4]]}];
```

```
v1 // MatrixForm
v2 // MatrixForm
v3 // MatrixForm
v4 // MatrixForm
```

```
vtilde1 // MatrixForm // N
vtilde2 // MatrixForm // N
vtilde3 // MatrixForm // N
vtilde4 // MatrixForm // N
```

Out[]//MatrixForm=

$$\begin{pmatrix} i \\ 0 \\ -1 \\ 1 \end{pmatrix}$$

Out[]//MatrixForm=

$$\begin{pmatrix} -i \\ 0 \\ -1 \\ 1 \end{pmatrix}$$

Out[]//MatrixForm=

$$\begin{pmatrix} 2 \\ -1 \\ 1 \\ 0 \end{pmatrix}$$

Out[]//MatrixForm=

$$\begin{pmatrix} -9 \\ 5 \\ -2 \\ 2 \end{pmatrix}$$

Out[]//MatrixForm=

$$\begin{pmatrix} 0.191206 + 1.77394 \, i \\ 0.287534 - 2.35959 \, i \\ -0.709973 - 0.594141 \, i \\ 1. \end{pmatrix}$$

Out[]//MatrixForm=

$$\begin{pmatrix} 0.191206 - 1.77394 \, i \\ 0.287534 + 2.35959 \, i \\ -0.709973 + 0.594141 \, i \\ 1. \end{pmatrix}$$

Out[]//MatrixForm=

$$\begin{pmatrix} 1.72126 \\ -0.348722 \\ -0.788653 \\ 1. \end{pmatrix}$$

Out[]//MatrixForm=

$$\begin{pmatrix} 1.22901 \\ -0.0135712 \\ -0.523043 \\ 1. \end{pmatrix}$$

In[]:= **(*Extract the eigenvalues*)**

$\lambda_1 = \text{eigA}[[1]]$

$\lambda_2 = \text{eigA}[[2]]$

$\lambda_3 = \text{eigA}[[3]]$

$\lambda_4 = \text{eigA}[[4]]$

Out[]:= **$3 + i$**

Out[]:= **$3 - i$**

Out[]:= **2**

Out[]:= **1**

We can verify that these are a valid set of eigenvectors by ensuring they satisfy

$$A \bar{V}_i = \lambda_i \bar{V}_i$$

```
In[ ]:= (*Verify  $\bar{v}_i = \lambda_i \bar{v}_i$ *)
```

```
A.v1 ==  $\lambda_1$  * v1
```

```
A.v2 ==  $\lambda_2$  * v2
```

```
A.v3 ==  $\lambda_3$  * v3
```

```
A.v4 ==  $\lambda_4$  * v4
```

```
Atilde.vtilde1 ==  $\lambda_1$  * vtilde1
```

```
Atilde.vtilde2 ==  $\lambda_2$  * vtilde2
```

```
Atilde.vtilde3 ==  $\lambda_3$  * vtilde3
```

```
Atilde.vtilde4 ==  $\lambda_4$  * vtilde4
```

```
Out[ ]:= True
```

```
Out[ ]:= True
```

```
Out[ ]:= True
```

```
Out[ ]:= True
```

```
Out[ ]:= True
```

```
Out[ ]:= True
```

```
Out[ ]:= True
```

```
Out[ ]:= True
```

We can now verify that $\text{eigenvectors}(\tilde{A}) = T^{-1} \text{eigenvectors}(A)$

```
In[ ]:= Inverse[T].v1 // MatrixForm // N
```

```
vtilde1 // MatrixForm // N
```

```
Out[ ]//MatrixForm=
```

$$\begin{pmatrix} 0.295266 - 0.102074 i \\ -0.413958 + 0.0432954 i \\ -0.0736073 + 0.144426 i \\ -0.0391457 - 0.170666 i \end{pmatrix}$$

```
Out[ ]//MatrixForm=
```

$$\begin{pmatrix} 0.191206 + 1.77394 i \\ 0.287534 - 2.35959 i \\ -0.709973 - 0.594141 i \\ 1. \end{pmatrix}$$

At first glance, it appears that $T^{-1} \bar{v}_1 \neq \tilde{v}_1$ but we can show that it satisfies $\tilde{A} T^{-1} \bar{v}_1 = \lambda_1 T^{-1} \bar{v}_1$ thereby showing it is an eigenvector associated with λ_1

```
In[ ]:= Atilde.Inverse[T].v1 == λ1 * Inverse[T].v1
Atilde.Inverse[T].v2 == λ2 * Inverse[T].v2
Atilde.Inverse[T].v3 == λ3 * Inverse[T].v3
Atilde.Inverse[T].v4 == λ4 * Inverse[T].v4
```

```
Out[ ]:= True
```

```
Out[ ]:= True
```

```
Out[ ]:= True
```

```
Out[ ]:= True
```

Property 4: Same trace

```
In[ ]:= TrA = Tr[A]
TrAtilde = Tr[Atilde]
TrA == TrAtilde
```

```
Out[ ]:= 9
```

```
Out[ ]:= 9
```

```
Out[ ]:= True
```

Property 5: Same rank

```
In[ ]:= rankA = MatrixRank[A]
rankAtilde = MatrixRank[Atilde]
rankA == rankAtilde
```

```
Out[ ]:= 4
```

```
Out[ ]:= 4
```

```
Out[ ]:= True
```

Diagonalization

We can use the modal transformation of $M = \text{eigenvectors}(A)$ to diagonalize the matrix using $M^{-1} A M$

```
In[ ]:= M = eigVecA;
Inverse[M].A.M // MatrixForm
eigA
```

```
Out[ ]//MatrixForm=
```

$$\begin{pmatrix} 3 + i & 0 & 0 & 0 \\ 0 & 3 - i & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

```
Out[ ]:= {3 + i, 3 - i, 2, 1}
```

Example 2: Defective Matrix

Consider the following defective matrix (meaning it cannot be diagonalized). We use the same matrix as before except we change the (1, 1) element ($3 \rightarrow 5$)

$$\text{In[]:= } \mathbf{A} = \begin{pmatrix} 5 & 4 & 2 & 1 \\ 0 & 1 & -1 & -1 \\ -1 & -1 & 3 & 0 \\ 1 & 1 & -1 & 2 \end{pmatrix};$$

We can use the same transformation matrix as before and calculate $\tilde{\mathbf{A}}$

```
In[ ]:= Atilde = Inverse[T].A.T;
Atilde // MatrixForm
Atilde // MatrixForm // N
```

Out[]//MatrixForm=

$$\begin{pmatrix} \frac{220365}{71732} & \frac{724585}{753186} & -\frac{156591}{35866} & -\frac{288913}{71732} \\ -\frac{230279}{143464} & \frac{1669453}{645588} & -\frac{61796}{17933} & -\frac{61537}{430392} \\ \frac{30175}{53799} & \frac{35090}{125531} & \frac{96483}{17933} & \frac{8760}{17933} \\ -\frac{118705}{107598} & \frac{1194281}{3389337} & -\frac{382507}{53799} & -\frac{12329}{322794} \end{pmatrix}$$

Out[]//MatrixForm=

$$\begin{pmatrix} 3.07206 & 0.962027 & -4.366 & -4.02767 \\ -1.60513 & 2.58594 & -3.44594 & -0.142979 \\ 0.560884 & 0.279533 & 5.38019 & 0.488485 \\ -1.10323 & 0.352364 & -7.10993 & -0.0381946 \end{pmatrix}$$

We can verify the aforementioned properties.

Property 1: Same determinant

```
In[ ]:= detA = Det[A]
detAtilde = Det[Atilde]
detA == detAtilde
```

Out[]:= 32

Out[]:= 32

Out[]:= True

Property 2: Same characteristic equations (and therefore same eigenvalues)

We can verify that $|A - \lambda I| = |\tilde{A} - \lambda I|$

```
In[ ]:= charEqA = Det[A - λ * IdentityMatrix[4]]
charEqAtilde = Det[Atilde - λ * IdentityMatrix[4]]
charEqA == charEqAtilde
```

Out[]:= $32 - 64\lambda + 42\lambda^2 - 11\lambda^3 + \lambda^4$

Out[]:= $32 - 64\lambda + 42\lambda^2 - 11\lambda^3 + \lambda^4$

Out[]:= True

Which means the eigenvalues are the same

```
In[ ]:= eigA = Eigenvalues[A]
eigAtilde = Eigenvalues[Atilde]
eigA == eigAtilde
```

```
Out[ ]:= {4, 4, 2, 1}
```

```
Out[ ]:= {4, 4, 2, 1}
```

```
Out[ ]:= True
```

Property 3: Similar eigenvectors

We can now verify that $\text{eigenvectors}(\tilde{A}) = T^{-1} \text{eigenvectors}(A)$

```
In[ ]:= (*Mathematica's Eigenvectors returns eigenvectors in row format,
transpose to get into col format*)
eigVecA = Transpose[Eigenvectors[A]];
eigVecAtilde = Transpose[Eigenvectors[Atilde]];
```

```
eigVecA // MatrixForm
eigVecAtilde // MatrixForm // N
```

```
Out[ ]//MatrixForm=
```

$$\begin{pmatrix} 1 & 0 & 1 & -1 \\ 0 & 0 & -1 & 1 \\ -1 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 \end{pmatrix}$$

```
Out[ ]//MatrixForm=
```

$$\begin{pmatrix} -0.920784 & 0. & 1.16923 & 1.0804 \\ 1.76664 & 0. & 0.979668 & 0.562692 \\ -0.337537 & 0. & -0.419543 & -0.285776 \\ 1. & 0. & 1. & 1. \end{pmatrix}$$

```
In[ ]:= (*Check that we extracted the eigenvectors correctly*)
```

```
v1 = Transpose[{eigVecA[[All, 1]]}];
```

```
v2 = Transpose[{eigVecA[[All, 2]]}];
```

```
v3 = Transpose[{eigVecA[[All, 3]]}];
```

```
v4 = Transpose[{eigVecA[[All, 4]]}];
```

```
vtilde1 = Transpose[{eigVecAtilde[[All, 1]]}];
```

```
vtilde2 = Transpose[{eigVecAtilde[[All, 2]]}];
```

```
vtilde3 = Transpose[{eigVecAtilde[[All, 3]]}];
```

```
vtilde4 = Transpose[{eigVecAtilde[[All, 4]]}];
```

```
v1 // MatrixForm
```

```
v2 // MatrixForm
```

```
v3 // MatrixForm
```

```
v4 // MatrixForm
```

```
vtilde1 // MatrixForm // N
```

```
vtilde2 // MatrixForm // N
```

```
vtilde3 // MatrixForm // N
```

```
vtilde4 // MatrixForm // N
```

```
Out[ ]//MatrixForm=
```

$$\begin{pmatrix} 1 \\ 0 \\ -1 \\ 1 \end{pmatrix}$$

```
Out[ ]//MatrixForm=
```

$$\begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$

```
Out[ ]//MatrixForm=
```

$$\begin{pmatrix} 1 \\ -1 \\ 0 \\ 1 \end{pmatrix}$$

```
Out[ ]//MatrixForm=
```

$$\begin{pmatrix} -1 \\ 1 \\ 0 \\ 0 \end{pmatrix}$$

```
Out[ ]//MatrixForm=
```

$$\begin{pmatrix} -0.920784 \\ 1.76664 \\ -0.337537 \\ 1. \end{pmatrix}$$

Out[]//MatrixForm=

$$\begin{pmatrix} 0. \\ 0. \\ 0. \\ 0. \end{pmatrix}$$

Out[]//MatrixForm=

$$\begin{pmatrix} 1.16923 \\ 0.979668 \\ -0.419543 \\ 1. \end{pmatrix}$$

Out[]//MatrixForm=

$$\begin{pmatrix} 1.0804 \\ 0.562692 \\ -0.285776 \\ 1. \end{pmatrix}$$

In[]:= **(*Extract the eigenvalues*)**

$\lambda_1 = \text{eigA}[[1]]$

$\lambda_2 = \text{eigA}[[2]]$

$\lambda_3 = \text{eigA}[[3]]$

$\lambda_4 = \text{eigA}[[4]]$

Out[]:= 4

Out[]:= 4

Out[]:= 2

Out[]:= 1

We can verify that these are a valid set of eigenvectors by ensuring they satisfy

$$A \bar{V}_i = \lambda_i \bar{V}_i$$


```
In[ ]:= (*Verify  $A \bar{v}_i = \lambda_i \bar{v}_i$ *)
```

```
A.v1 ==  $\lambda_1 * v_1$ 
```

```
A.v2 ==  $\lambda_2 * v_2$ 
```

```
A.v3 ==  $\lambda_3 * v_3$ 
```

```
A.v4 ==  $\lambda_4 * v_4$ 
```

```
Atilde.vtilde1 ==  $\lambda_1 * vtilde1$ 
```

```
Atilde.vtilde2 ==  $\lambda_2 * vtilde2$ 
```

```
Atilde.vtilde3 ==  $\lambda_3 * vtilde3$ 
```

```
Atilde.vtilde4 ==  $\lambda_4 * vtilde4$ 
```

```
Out[ ]:= True
```

```
Out[ ]:= True
```

```
Out[ ]:= True
```

```
Out[ ]:= True
```

```
Out[ ]:= True
```

```
Out[ ]:= True
```

```
Out[ ]:= True
```

```
Out[ ]:= True
```

We can now verify that $\text{eigenvectors}(\tilde{A}) = T^{-1} \text{eigenvectors}(A)$

```
In[ ]:= Inverse[T].v1 // MatrixForm // N
```

```
vtilde1 // MatrixForm // N
```

```
Out[ ]//MatrixForm=
```

$$\begin{pmatrix} 0.193191 \\ -0.370662 \\ 0.0708192 \\ -0.209812 \end{pmatrix}$$

```
Out[ ]//MatrixForm=
```

$$\begin{pmatrix} -0.920784 \\ 1.76664 \\ -0.337537 \\ 1. \end{pmatrix}$$

At first glance, it appears that $T^{-1} \bar{v}_1 \neq \tilde{v}_1$ but we can show that it satisfies $\tilde{A} T^{-1} \bar{v}_1 = \lambda_1 T^{-1} \bar{v}_1$ thereby showing it is an eigenvector associated with λ_1

```
In[ ]:= Atilde.Inverse[T].v1 == λ1 * Inverse[T].v1
        Atilde.Inverse[T].v2 == λ2 * Inverse[T].v2
        Atilde.Inverse[T].v3 == λ3 * Inverse[T].v3
        Atilde.Inverse[T].v4 == λ4 * Inverse[T].v4
```

```
Out[ ]:= True
```

```
Out[ ]:= True
```

```
Out[ ]:= True
```

```
Out[ ]:= True
```

Property 4: Same trace

```
In[ ]:= TrA = Tr[A]
        TrAtilde = Tr[Atilde]
        TrA == TrAtilde
```

```
Out[ ]:= 11
```

```
Out[ ]:= 11
```

```
Out[ ]:= True
```

Property 5: Same rank

```
In[ ]:= rankA = MatrixRank[A]
        rankAtilde = MatrixRank[Atilde]
        rankA == rankAtilde
```

```
Out[ ]:= 4
```

```
Out[ ]:= 4
```


```
Out[ ]:= True
```

Diagonalization

In this situation we cannot use the modal transformation of $M = \text{eigenvectors}(A)$ to diagonalize the matrix because the eigenvectors do not span the space, meaning that M is singular so M^{-1} does not exist

```
In[ ]:= M = eigVecA;
        MatrixRank[M]
        Inverse[M]
```

```
Out[ ]:= 3
```

 **Inverse:** Matrix $\{\{1, 0, 1, -1\}, \{0, 0, -1, 1\}, \{-1, 0, 0, 0\}, \{1, 0, 1, 0\}\}$ is singular.

```
Out[ ]:= Inverse[{{1, 0, 1, -1}, {0, 0, -1, 1}, {-1, 0, 0, 0}, {1, 0, 1, 0}}]
```

This can be ameliorated using Jordan Normal form (https://en.wikipedia.org/wiki/Jordan_normal_form) but this is a topic for another video.