

The following notes are useful for this discussion: [Note 13](#), [Note 15](#).

### 1. Towards Upper-Triangularization By An Orthonormal Basis

Previously in this course, we have seen the value of changing our coordinates to be eigenbasis-aligned, because we can then view the system as a set of parallel scalar systems. Diagonalization causes these scalar equations to be fully uncoupled such that they can be solved separately. But even when we cannot diagonalize, we can *upper-triangularize* such that we can still solve the equations one at a time, from the "bottom up".

To better understand the steps involved, we will use the following concrete example:

$$M = S_{[3 \times 3]} = \begin{bmatrix} \frac{5}{12} & \frac{5}{12} & \frac{1}{6} \\ \frac{5}{12} & \frac{5}{12} & \frac{1}{6} \\ \frac{1}{6} & \frac{1}{6} & \frac{2}{3} \end{bmatrix} \quad (1)$$

and solve the general case by abstracting variables. Note that there is a datahub link to a jupyter notebook on the website, which will allow you to perform the numerical calculations quickly.<sup>1</sup>

- (a) Consider a non-zero vector  $\vec{u}_0 \in \mathbb{R}^n$ . Can you think of a way to extend it to a set of basis vectors for  $\mathbb{R}^n$ ? In other words, find  $\vec{u}_1, \dots, \vec{u}_{n-1}$ , such that  $\text{span}(\vec{u}_0, \vec{u}_1, \dots, \vec{u}_{n-1}) = \mathbb{R}^n$ . **To make things concrete, consider**  $\begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}$ . **Can you get an orthonormal basis where the first vector is a multiple of this vector?**

*(HINT: What was the last discussion all about? Also, the given vector isn't normalized yet!)*

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<sup>1</sup>This particular matrix has an additional special property of symmetry, but we won't be invoking that here.

- (b) Now consider a real eigenvalue  $\lambda_1$ , and the corresponding (normalized) eigenvector  $\vec{v}_1 \in \mathbb{R}^n$  of  $M \in \mathbb{R}^{n \times n}$  ( $M\vec{v}_1 = \lambda_1\vec{v}_1$ ). We know we can extend  $\vec{v}_1$  to an orthonormal basis of  $\mathbb{R}^n$ . We will denote the basis by

$$U = \begin{bmatrix} | & | & \cdots & | \\ \vec{u}_1 & \vec{u}_2 & \cdots & \vec{u}_n \\ | & | & \cdots & | \end{bmatrix} \quad (2)$$

where  $\vec{u}_1 = \vec{v}_1$  (note that this eigenvector is already normalized).

Our goal is to look at what the matrix  $M$  looks like in the coordinate system defined by the

basis  $U$ . **Compute  $U^T M U$  by writing  $U = [\vec{v}_1 \ R]$ , where  $R := \begin{bmatrix} | & | & \cdots & | \\ \vec{r}_1 & \vec{r}_2 & \cdots & \vec{r}_{n-1} \\ | & | & \cdots & | \end{bmatrix}$ .** (Note :

$\vec{r}_i = \vec{u}_{i+1}$ )

- (c) **Verify that  $U^{-1} = U^T$ , where  $U$  is the matrix we get from Gram-Schmidt process.**

(d) Look at the first column and the first row of  $U^T M U$  and show that:

$$M = U \begin{bmatrix} \lambda_1 & \vec{a}^T \\ \vec{0} & Q \end{bmatrix} U^T \quad (3)$$

where  $Q = R^T M R$ . Here,  $\vec{a}$  is a vector related to  $M$ ,  $R$ , and  $\vec{v}_1$  (we will show the relation!).

(e) Now, we can recurse on  $Q$  to get:

$$Q = \begin{bmatrix} \vec{v}_2 & Y \end{bmatrix} \begin{bmatrix} \lambda_2 & \vec{b}^T \\ \vec{0} & P \end{bmatrix} \begin{bmatrix} \vec{v}_2 & Y \end{bmatrix}^T \quad (4)$$

where we have taken  $\vec{v}_2 \in \mathbb{R}^{n-1}$ , a normalized eigenvector of  $Q$ , associated with eigenvalue  $\lambda_2$ . Again  $\vec{v}_2$  is extended into an orthonormal basis to form  $\begin{bmatrix} \vec{v}_2 & Y \end{bmatrix}$ .

**Plug this form of  $Q$  into  $M$  above, to show that:**

$$M = \begin{bmatrix} \vec{v}_1 & R\vec{v}_2 & RY \end{bmatrix} \begin{bmatrix} \lambda_1 & \check{a}_1 & \check{\vec{a}}_{\text{rest}}^T \\ 0 & \lambda_2 & \vec{b}^T \\ \vec{0} & \vec{0} & P \end{bmatrix} \begin{bmatrix} \vec{v}_1 & R\vec{v}_2 & RY \end{bmatrix}^T \quad (5)$$

where we define  $\check{\vec{a}}$  to be the "adjusted"  $\vec{a}$  to account for the substitution of  $Q$ ;  $\check{\vec{a}}^T = \vec{a}^T \begin{bmatrix} \vec{v}_2 & Y \end{bmatrix}$ .

(f) **(PRACTICE)** Show that the matrix  $\begin{bmatrix} \vec{v}_1 & R\vec{v}_2 & RY \end{bmatrix}$  is still orthonormal.

(g) **(PRACTICE)** We have shown how to upper triangularize a  $3 \times 3$  and a  $2 \times 2$  matrix. **How can we generalize this process to any  $n \times n$  matrix  $M$ ?**

- (h) **(PRACTICE)** Show that the characteristic polynomial of square matrix  $M$  is the same as that of the square matrix  $UMU^{-1}$  for any invertible  $U$ . You should use the key property  $\det(AB) = \det(A)\det(B)$  for square matrices.

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