

**Matrix Theory**  
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**Partitioning the SVD**

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Recall: For  $A \in \mathbb{C}^{n \times n}$ ,  $\|A\|_2 = \sqrt{\mu}$ , where  $\mu = \max \text{Eigenvalue of } A^*A$ .

Partitioning the SVD:

$A \in \mathbb{C}^{m \times n}$  has  $r \leq p \triangleq \min(m, n)$  nonzero SVals and  $p-r$  zero SVals.

$$\sigma_1 \geq \dots \geq \sigma_r > \sigma_{r+1} = \dots = \sigma_p = 0.$$

$$A = U \Sigma V^H$$


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$p-r$  zero SVals.

$$\sigma_1 \geq \dots \geq \sigma_r > \sigma_{r+1} = \dots = \sigma_p = 0.$$

$$A = U \Sigma V^H$$

$$= \begin{bmatrix} u_1 & u_2 \end{bmatrix}_{\substack{n \\ m-n}} \begin{bmatrix} \tilde{\Sigma} & 0 \\ 0 & 0 \end{bmatrix}_{\substack{n \times n \\ m \times n}} \begin{bmatrix} v_1^H \\ v_2^H \end{bmatrix}_{\substack{\leftarrow n \\ \leftarrow n-n}} \quad (*)$$

where  $\tilde{\Sigma} = \text{diag}(\sigma_1 \dots \sigma_r)$  ( $r \times r$  matrix).


So, there is one other thing I want to make say, which is that this is about partitioning. This so this partitioning actually brings out the core structure of the matrix and what the singular value decomposition allows you to do is to partition matrices as I am going to tell you now. So, here it is.

So, again  $A$  is in  $\mathbb{C}$  to the  $m$  by  $n$  and has so suppose this has  $r$  which is at most  $p$  defined to be  $\min$  of  $m, n$  non-zero singular values and say  $p - r$  zero singular values. So, we will use the notation  $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_r > 0$ , which is greater than  $\sigma_{r+1} = \dots = \sigma_p = 0$ . Then I can write.

So, since  $u$  is unitary  $u^H u = I$  Hermitian  $Av$  equals  $\sigma$  can be written as this is equal to  $u \sigma v^H$  Hermitian, which I can partition like this  $u_1 \ u_2$  where this is the first  $r$  columns, this is the next  $m - r$  columns and here I have say  $\tilde{\sigma}$  or  $D$  as I called it in this stating the theorem 0.0. This is overall an  $m$  by  $n$  matrix and this itself is  $r$  cross  $r$  and  $v_1 \ v_2$  Hermitian and this has the first  $r$  rows and this has the next  $n - r$  rows.

We will call this form star where  $\tilde{\sigma}$  is equal to a diagonal matrix containing  $\sigma_1$  through  $\sigma_r$  on the diagonal. So, this allows you to partition the singular value decomposition like this into blocks and what this partitioning does is it reveals a lot of structure in this matrix.

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where  $\tilde{\Sigma} = \text{diag}(\sigma_1 \dots \sigma_r)$  ( $r \times r$  matrix).

$$1. \text{rank}(A) = r$$

$$A = [u_1 \ u_2] \begin{bmatrix} \tilde{\Sigma} & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} v_1^H \\ v_2^H \end{bmatrix} = u_1 \tilde{\Sigma} v_1^H$$

"Economy SVD"

$$= u_1 B, \quad B \triangleq \tilde{\Sigma} v_1$$

$\Rightarrow A$  has exactly  $r$  l.i. cols.

$$2. \mathcal{N}(A) = \mathcal{R}(v_2), \text{ moreover, } v_2 \text{ is a basis for } \mathcal{N}(A).$$

$V_1 \perp V_2, \quad V_2^H V_2 = I_{n-r}$   
 $Ax = [u_1, u_2] \begin{bmatrix} \tilde{\Sigma} & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} v_1^H \\ v_2^H \end{bmatrix} v_2 c$   
 $= [u_1, u_2] \begin{bmatrix} \tilde{\Sigma} & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 0 \\ c \end{bmatrix} = 0$   
 $\text{Span}(V_2) \subseteq \mathcal{N}(A).$   
 However if  $x$  contains components of  $V_1$  then  
 $Ax \neq 0$  by the above eqn. itself.  
 Since  $[V_1, V_2]$  span  $\mathbb{C}^n \Rightarrow V_2$  is a basis for  $\mathcal{N}(A)$ .

So, first of all, it shows that the rank of  $A$  is exactly equal to  $r$  and you can write  $A$  as so if I do this multiplication here, I get  $\tilde{\Sigma} V_1^H$  Hermitian and then this whole thing multiplies 0 down here and so I can write this like this  $U_1 U_2 \tilde{\Sigma}$ .

So, only  $V_1$  Hermitian is of size  $r$  that only multiplies this, whereas this does not this is just multiplying the, if I take a particular column here, the entries in  $V_2$  Hermitian are all multiplying 0s here, so they do not contribute to anything. So, this can be written as this times  $V_1$  Hermitian and then down here, I will have all the 0s and this itself is multiplying  $U_2$  here. So, this does not contribute anything. So, I can write this as  $U_1 \tilde{\Sigma} V_1^H$  and this is called the economy SVD. Because this thing here is now an  $r$  cross  $r$  matrix and this is of  $m$  by  $r$  and this is  $r$  by  $n$ .

And so I can see that I see that now that  $A$  is equal to  $U_1$  times the matrix  $B$ , where  $B$  is  $\tilde{\Sigma} V_1^H$ . So, what that means is that the  $i$ th column of  $A$  is a linear combination of the  $r$  columns of  $U_1$  with coefficients given by the  $i$ th column of  $B$ . So, that the  $i$ th column  $B$  gives the coefficients with which I should combine the columns of  $U_1$  to get the  $i$ th column of  $A$ .

So, and this  $\tilde{\Sigma}$ , of course, is diagonal with non-zero entries and so it has linearly independent columns and  $V_1$  Hermitian has linearly independent rows and so  $\tilde{\Sigma} V_1^H$  or this matrix  $B$  has linearly independent rows and so what this means is that  $A$  has exactly  $r$  linearly independent columns and that is why the rank of  $A$  equals  $r$ . The second property is that

the null space of  $A$  is the same as the span of the columns of  $v_2$  and further  $v_2$  is a basis for the null space of  $A$ . So, null space of  $A$  is the set of all vectors  $x$  such that  $Ax$  equals  $0$ .

Now, if  $x$  is in the span of  $v_2$ , then I can write  $x$  to be equal to  $v_2$  times  $c$  where  $c$  is in some or I should be more careful  $n$  minus  $r$ . So it is in the span of  $v_2$ . Now,  $v_1$  is perpendicular to  $v_2$ , because the columns of  $v$  form an orthonormal set the matrix  $v$  itself is a unitary matrix. So,  $v_1$  is perpendicular to  $v_2$  and  $v_2^H v_2$  is the identity matrix of size  $n$  minus  $r$ .

So, then what I have from this is that, if I take  $Ax$ , this is equal to  $u_1, u_2$  times  $\sigma_1, \sigma_2, \dots, \sigma_r$  times  $v_1^H v_2^H$  times  $v_2 c$ . Now,  $v_1^H v_2$  is  $0$ ,  $v_2^H v_2$  is the identity matrix. But then this is only going to multiply these  $0$  entries here and so what I will be left with is  $u_1, u_2$  times  $\sigma_1, \sigma_2, \dots, \sigma_r$ . And if I take this product first I get  $0, v_2^H v_2$  is the identity matrix.

So, I will get  $c$  down here and so now, if I expand this out, I will get  $0$  times  $\sigma_1, \sigma_2, \dots, \sigma_r$  and then  $0$  is down here. So, this whole thing becomes equal to the  $0$  vector. So, of  $v_2$  is a subset of the null space of  $A$  for any vector that belongs the span of  $v_2$  satisfies  $Ax$  equals  $0$ . So, it is a subset of the null space of  $A$ .

However, if  $x$  contains components of  $v_1$ , that is to say that if the projection of  $x$  on to span of  $v_1$  is nonzero, then  $Ax$  cannot be equal to  $0$  by this equation itself because if it has some component of  $v_1$  then  $v_1^H$  times that will be a non-zero quantity. So, I will get a non-zero quantity up here and that is multiplying a non-zero  $\sigma_i$  which is in turn multiplying a non-zero  $u_i$  by the above equation itself and  $v_1, v_2$  form a complete basis for  $\mathbb{C}^n$  together they span  $\mathbb{C}^n$  which implies that  $v_2$  is basis for the null space of  $A$ .

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However if  $z$  contains components of  $v_1$  then  
 $Ax \neq 0$  by the above eqn. itself.  
Since  $[v_1, v_2]$  span  $\mathbb{C}^n \Rightarrow v_2$  is a basis for  $N(A)$ .

3.  $\mathcal{R}(A) = \mathcal{R}(u_1)$   
4.  $\mathcal{R}(A^H) = \mathcal{R}(v_1)$   
5.  $\mathcal{R}(A)_\perp = N(A^H) = \mathcal{R}(u_2)$

More properties...

So, we can similarly say many more things like the range space of  $A$  is the range space of  $u_1$ . The range space of  $A$  Hermitian is the range space of  $v_1$  and the orthogonal complement of the range space of  $A$  which is equal to the null space of  $A$  Hermitian is equal to the range space of  $u_2$  and so on and we will see some more properties. So, we are out of time, so more properties. So, we will see that in the next class. So, that is all I have for today. We will continue again on Wednesday.