# The rank nullity theorem 

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Let $A$ be an $n \times m$ matrix. We would like to use Gauss elimination to find a basis for the column space of $A$. In order to do this we need to understand more about the relationship between $A$ and its reduced row echelon form.

Definition 1 Two matrices $A$ and $A^{\prime}$ are said to be row equivalent if there exists a sequence of elementary row operations which transforms $A$ to $A^{\prime}$.

Note that $A$ is row equivalent to itself. Furthermore, if $A$ is row equivalent to $A^{\prime}, A^{\prime}$ is row equivalent to $A$, because elementary row operations are reversible. Finally, note that if $A$ is row equivalent to $A^{\prime}$ and $A^{\prime}$ is row equivalent to $A^{\prime \prime}$, then $A$ is row equivalent to $A^{\prime \prime}$, since we can chain together the necessary sequences of elementary row operations.

Proposition 2 Let $A$ and $A^{\prime}$ be $n \times m$ matrices. Then $A$ and $A^{\prime}$ are row equivalent if and only if there exists an invertible $n \times n$ matrix $U$ such that $A^{\prime}=U A$.

Let me just explain how to find $U$. (In fact there can be many such $U$.) This is similar to the algorithm we used for inverting a matrix, only here we do not require that $A$ be square. Just form the $n \times(m+n)$ matrix $\left(A \mid I_{n}\right)$, and apply the elementary row operations necessary to change $A$ to $A^{\prime}$. The matrix you get is $\left(A^{\prime} \mid U\right)$, and $U A=A^{\prime}$. This is easy to see if $A^{\prime}$ can be obtained by a single elementary row operation just by checking, and the general case follows by chaining.

Corollary 3 If $A^{\prime}$ and $A$ are row equivalent, then

1. $A^{\prime}$ and $A$ have the same null space.
2. $A^{\prime}$ and $A$ have the same row space (the span of the rows).

Proof: Actually we have already verified the first fact. As an exercise, deduce it from proposition 2. The second fact is easily seen if $A^{\prime}$ is obtained from $A$ by a single elementary row operation just from the definitions, and the general case follows by a chaining argument.

It is now possible to prove that the reduced row echelon form of a matrix is unique. Look for the note on this on the web page. Our main goal here is to show how to use Gauss elimination to find a basis for the column space of a matrix. The difficulty is that the column space (unlike the row space) changes when elementary row operations are performed.

Theorem 4 Let $A$ be an $n \times m$ matrix and let $A^{\prime}$ be its reduced row echelon form.

1. The null space of $A$ is equal to the null space of $A^{\prime}$.
2. The rank $r$ of $A$ is equal to the rank of $A^{\prime}$.
3. If $\left(\ell_{1}, \ldots \ell_{r}\right)$ are the leading indices of $A^{\prime}$, then the corresponding columns ( $\left.C_{\ell_{1}}\left(A^{\prime}\right), \ldots C_{\ell_{r}}\left(A^{\prime}\right)\right)$ forms a basis for the column space of $A^{\prime}$, and
4. $\left(C_{\ell_{1}}(A), \ldots C_{\ell_{r}}(A)\right)$ forms a basis for the column space of $A$.

The third statement is easy to see by inspection of the reduced row echelon matrix $A$. The last statement is not explained well in the book. It is easy to understand it, using the proposition, since $A=U A^{\prime}$ for some invertible matrix $U$. For any $j, C_{j}\left(A^{\prime}\right)$ can be written as a linear combination of the leading index columns of $A^{\prime}$. Then

$$
\begin{aligned}
C_{j}\left(A^{\prime}\right) & =x_{1} C_{\ell_{1}}\left(A^{\prime}\right)+\cdots x_{r} C_{\ell_{r}}\left(A^{\prime}\right) \\
U C_{j}\left(A^{\prime}\right) & =U\left(x_{1} C_{\ell_{1}}\left(A^{\prime}\right)+\cdots x_{r} C_{\ell_{r}}\left(A^{\prime}\right)\right) \\
& =x_{1} U C_{\ell_{1}}\left(A^{\prime}\right)+\cdots x_{r} U C_{\ell_{r}}\left(A^{\prime}\right) \\
C_{j}\left(U A^{\prime}\right) & \left.\left.=x_{1} C_{\ell_{1}} U A^{\prime}\right)+\cdots x_{r} C_{\ell_{r}} U A^{\prime}\right) \\
C_{j}(A) & =x_{1} C_{\ell_{1}}(A)+\cdots x_{r} C_{\ell_{r}}(A)
\end{aligned}
$$

This shows that every column of $A$ is a linear combination of the leading index columns, so they span the column space. A similar argument, using $U^{-1}$ shows that there are no redundancies among these columns, so they form a basis for the column space of $A$.

