

## Overview: row-space solution to $Ax = b$

This method begins with the realization that it's possible to effect linear operations on the rows of a matrix by left multiplying it with a suitable matrix. By extension, it's possible to perform a Gram-Schmidt orthonormalization of the rows of a matrix through such repeated operations.

In the case at hand, the system of linear equations is described by the matrix equation  $Ax = b$ , where  $A$  is the  $m$ -by- $n$  coefficient matrix,  $x$  the  $n$ -dimensional vector of unknowns, and  $b$  is an  $m$ -dim vector. Also, take  $m \leq n$ , assume  $A$  has full row rank (the general case is treated in [1]), and assume all entries in  $A, b$  are real.

Determine a sequence of row operations, implemented by a sequence of  $m$ -by- $m$  matrices  $M_k$  that orthonormalize the rows of  $A$ , and apply them from the left to the original equation to obtain

$$(\cdots M_2 M_1)Ax = (\cdots M_2 M_1)b \quad (1)$$

or  $A'x = b'$ , where  $A' = MA$ ,  $b' = Mb$ , and  $M = (\cdots M_2 M_1)$ . (Keep in mind that the matrices  $M_s$  do not need to be *explicitly* created; only their actions on the rows need to be implemented.) At this point the reader should pause and realize that  $x \in R^n$  and that there is an  $m$ -dimensional orthonormal basis available via the rows of  $A'$ . It follows that  $x$  may be expressed as

$$x = \alpha_1 v_1^T + \alpha_2 v_2^T + \cdots + \alpha_m v_m^T + \beta_1 w_1 + \cdots + \beta_{n-m} w_{n-m} \quad (2)$$

where the  $\alpha_i$  and  $\beta_j$  are suitable coefficients,  $v_i$  are the rows of  $A'$ ,  $w_j$  represents a basis for the null space,  $i = 1, \dots, m$ , and  $j = 1, \dots, n - m$ . A simpler way to write out this solution is just to realize that the  $v_i^T$  are columns of  $(A')^T$  and set

$$x = (A')^T \alpha + x_{null} \quad (3)$$

where  $\alpha = (\alpha_1, \dots, \alpha_m)^T$ , and  $x_{null}$  represents the previous sum  $\sum \beta_j w_j$ . (Another way to express  $x_{null}$  is  $Py$ , where  $P$  is the null space projection matrix  $(1 - A'^T A')$  and  $y$  is an arbitrary  $n$ -dimensional vector.) After substituting this expression into  $A'x = b'$ , it follows that  $\alpha = b'$ , and the solution may be written as

$$x = (A')^T b' + x_{null} \quad (4)$$

Finally, observe that since the particular solution is built from the row space (and not the null space) of  $A$ , it is a *minimum norm* solution. This also follows from a characterization of its corresponding generalized inverse (see [1]).

## References

- [1] M.F. Zimmer, Two direct solvers for a system of linear equations. *arXiv preprint*, arXiv:1611.06633, 2020.