Linear Equations for Subspaces, Cramer's Rule

The solution set of a linear system AX = 0 is a linear subspace of dimension n - r where *r* is the row-rank of *A*. If *A* is an $m \times n$ -matrix then AX = 0 stands for *m*-many homogeneous equations in *n*-many unknowns:

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x = 0$$

$$a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = 0$$

$$\dots$$

$$a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nn}x_n = 0$$

Each equation $a_{i1}x_1 + a_{i2}x_2 + ... + a_{in}x_n = 0$ stands for an n - 1 dimensional subspace of F^n assuming that not all a_{ij} are zero. The solutions of AX = 0 then is the intersection of m-many hyperplanes of F^n . We also know how to find a basis of the solution space U of AX = 0. Without loss of generality we may assume that the first unknowns $x_1, x_2, ..., x_r$ are expressed in terms of $x_{r+1}, x_{r+2}, ..., x_n$. Choosing for $x_{r+1}, x_{r+2}, ..., x_n$ successively 1's and zeroes we get a basis $X_{r+1}, X_{r+2}, ..., X_n$ of U.

The easiest case is one homogeneous equation in n –unknowns. Like

$$x_1 + a_2 x_2 + \dots + a_n x_n = 0, x_1 = -a_2 x_2 - \dots - a_n x_n$$
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and we choose successively for $x_2 = 1, x_3 = ... = x_n = 0$ then $x_2 = 0, x_3 = 1, ..., x_n = 0, x_2 = x_3 = 0, ..., x_n = 1$ and get n - 1 basis vectors $X_2, X_3, ..., X_n$ and U is the set

$$x_2X_2 + x_3X_3 + \ldots + x_nX_n$$

of all linear combinations of the X_i . For example a hyperplane of \mathbb{R}^4 is given by one linear equation in 4 unknowns, like

$$2x_1 + 4x_2 + 3x_3 + 6x_4 = 0 \qquad \#$$

then $x_1 = -2x_2 - 3/2x_3 - 3x_4$ and three basis vectors of the hyper plane are

$$X_{2} = \begin{pmatrix} -2 \\ 1 \\ 0 \\ 0 \end{pmatrix}, X_{3} = \begin{pmatrix} -3/2 \\ 0 \\ 1 \\ 0 \end{pmatrix}, X_{4} = \begin{pmatrix} -3 \\ 0 \\ 0 \\ 1 \end{pmatrix}$$

and all linear combinations $x_2X_2 + x_3X_3 + x_4X_4$ make up the hyper plane *U* Now, a subspace *U* of F^n may be given by *m*-many vectors α_1 of F^n . Thus $U = \langle \alpha_1, \alpha_2, \dots, \alpha_m \rangle$. If we define an $m \times n$ matrix by

$$A = \begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \cdots \\ \alpha_m \end{pmatrix}$$

then elementary row operations don't change the row space, that is U. If dim U = r then there are r –many non-zero rows of the row-echelon form of A from which we can read off n - r many basic solutions of AX = 0. This n - r –dimensional subspace of F^n is the space $U^{\perp} = \{X | \alpha_1 \perp X, \alpha_2 \perp X, \dots, \alpha_m \perp X\}$, recall that

$$(a_1, \alpha_2, \dots, \alpha_n) \begin{pmatrix} b_1 \\ b_2 \\ \dots \\ b_n \end{pmatrix} = a_1 b_1 + a_2 b_2 + \dots + a_n b_n = 0 \text{ iff } \alpha \perp \beta$$

Now let $\beta_1, \beta_2, \dots, \beta_{n-r}$ be a basis of AX = 0. That is of U^{\perp} . We choose these vectors as rows of a $(n - r) \times n$ -matrix

$$B = \begin{pmatrix} \beta_1 \\ \beta_2 \\ \dots \\ \beta_{n-r} \end{pmatrix}$$

Amongst the solutions of BX = 0 are all vectors α_i . Because $\alpha_i \cdot \beta_j = \beta_j \cdot \alpha_i = 0$. This is because the β_j are solutions of AX = 0. The α_i span the space U where dim U = r. The solution space of BX = 0 has dimension n - (n - r) = r. Thus BX = 0 has solution space U.

Determinants were useful to provide explicit formulas for solutions of linear equations. Nowadays only the special case of *n* –equations with *u* –unknowns is still important. We assume that in AX = B the rank of *A* is *n*. That is, for every choice of *B* there is a unique solution $X = A^{-1}B$. We can find *X* according to *Cramer's* Rule. Given is the square matrix *A* where *A* is of rank *n*. Then let $X = (x_1, x_2, ..., x_n)$ be the solution for AX = B. We have that

 $x_{1} \det(A) = \det \begin{pmatrix} a_{11}x_{1} & a_{12} & \dots & a_{1n} \\ a_{2} \cdot 1x_{11} & a_{22} & \dots & a_{2n} \\ \dots & \dots & \dots & \dots \\ a_{n1}x_{1} & a_{n2} & \dots & a_{nn} \end{pmatrix} = \det \begin{pmatrix} a_{11}x_{1} + a_{12}x_{2} + \dots + a_{1n}x_{n} & a_{12} & \dots & a_{1n} \\ a_{2} \cdot 1x_{11} + a_{22}x_{2} + \dots + a_{2n}x_{n} & a_{22} & \dots & a_{2n} \\ \dots & \dots & \dots & \dots & \dots \\ a_{n1}x_{1} + a_{n2}x_{2} + \dots + a_{nn}x_{n} & a_{n2} & \dots & a_{nn} \end{pmatrix} = \det \begin{pmatrix} a_{11}x_{1} + a_{12}x_{2} + \dots + a_{1n}x_{n} & a_{12} & \dots & a_{1n} \\ a_{2} \cdot 1x_{11} + a_{22}x_{2} + \dots + a_{2n}x_{n} & a_{22} & \dots & a_{2n} \\ \dots & \dots & \dots & \dots & \dots \\ a_{n1}x_{1} + a_{n2}x_{2} + \dots + a_{nn}x_{n} & a_{n2} & \dots & a_{nn} \end{pmatrix}$

which gives us

$$x_{1} = \frac{\det \begin{pmatrix} b_{1} & a_{12} & \dots & a_{1n} \\ b_{2} & a_{22} & \dots & a_{2n} \\ \dots & \dots & \dots & \dots \\ b_{n} & a_{2n} & \dots & a_{nn} \end{pmatrix}}{\det \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{2} & a_{22} & \dots & a_{2n} \\ \dots & \dots & \dots & \dots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{pmatrix}}$$

and similarly for the other x_i , i = 2, ... n.