What is the size of vector w and y?

the size of w is d+1 and the size of y is n because $w = \begin{bmatrix} w_0 \\ \vdots \\ w_d \end{bmatrix}$ and $y = \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix}$

What is the size of matrix A? Write A.

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A = \begin{bmatrix} 1 & x_1 & x_1^2 & x_1^3 & \cdots & x_1^d \\ 1 & x_2 & x_2^2 & x_2^3 & \cdots & x_2^d \\ 1 & x_3 & x_3^2 & x_3^3 & \cdots & x_3^d \\ 1 & x_4 & x_4^2 & x_4^3 & \cdots & x_1^d \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_n & x_n^2 & x_n^3 & \cdots & x_n^d \end{bmatrix} so that the size of matrix A is n \times (d+1)
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Let d + 1 = n, then, A becomes a square matrix. Compute the determinant of A

If there is a, b such that $a \neq b$ and $x_a = x_b$, then A has two same rows so that $\det A = 0$ Assume that there is no a, b such that $a \neq b$ and $x_a = x_b$

$$\begin{aligned} & \text{then, } |A| = \begin{vmatrix} 1 & x_1 & x_1^2 & x_2^3 & \cdots & x_1^{n-1} \\ 1 & x_2 & x_2^2 & x_2^3 & \cdots & x_2^{n-1} \\ 1 & x_4 & x_4^2 & x_4^3 & \cdots & x_2^{n-1} \\ 1 & x_4 & x_4^2 & x_4^3 & \cdots & x_2^{n-1} \\ 1 & x_4 & x_4^2 & x_4^3 & \cdots & x_2^{n-1} \\ 1 & x_4 & x_4^2 & x_4^3 & \cdots & x_2^{n-1} \\ 1 & x_4 & x_4^2 & x_4^3 & \cdots & x_2^{n-1} \\ 1 & x_4 & x_4^2 & x_4^3 & \cdots & x_2^{n-1} \\ 1 & x_4 & x_4^2 & x_4^3 & \cdots & x_2^{n-1} \\ 1 & x_4 & x_4^2 & x_4^3 & \cdots & x_2^{n-1} \\ 1 & x_4 & x_4^2 & x_4^3 & \cdots & x_2^{n-1} \\ 1 & x_4 & x_4^2 & x_4^3 & \cdots & x_2^{n-1} \\ 1 & x_4 & x_4^2 & x_4^2 & \cdots & x_2^{n-2} & x_4^{n-2} & x_4^{n-2} & x_4^{n-2} \\ 1 & x_4 & x_4^2 & x_4^2 & \cdots & x_2^{n-2} & x_4^{n-2} & x_4^{n-2} & x_4^{n-2} \\ 1 & x_4 & x_4^2 \\ 1 & x_4 & x_4^2 \\ 1 & x_4 & x_4^2 \\ 1 & x_4 & x_4^2 \\ 1 & x_4 & x_4^2 \\ 1 & x_4 & x_4^2 \\ 1 & x_4 & x_4^2 \\ 1 & x_4 & x_4^2 \\ 1 & x_4 & x_4^2 \\ 1 & x_4 & x_4^2 \\ 1 & x_4 & x_4^2 \\ 1 & x_4 & x_4^2 & x_$$

Through the above equation, I hypothesized that $\det A = \prod_{j=1}^{n-1} \prod_{i=j+1}^n (x_i - x_j)$ (for $n \ge 2$) [for n = 1, $\det A = 1$]

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Let P(n) denote 'det A = \prod_{j=1}^{n-1} \prod_{i=j+1}^{n} (x_i - x_j)' (basic step) P(2) \equiv \begin{vmatrix} 1 & x_1 \\ 1 & x_2 \end{vmatrix} = x_2 - x_1 \equiv T
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(Induction step) Assume that P(k) \equiv T for k \in \mathbb{N}, which means that  \begin{vmatrix} 1 & c_2 & \cdots & c_2^{k-1} \\ \vdots & \vdots & \ddots & \vdots \end{vmatrix} 
                                                                                                                                                                  =\prod_{j=1}^{k-1}\prod_{i=j+1}^{k}(c_i-c_j)
                                                                      x_2 \quad \cdots \quad x_2^{k-1}
\vdots \quad \ddots \quad \vdots
                                                x_2^k
                                                                                                      x_2^{k-1}(x_2-x_1)
                                                                                                                                     [ : ( n^{th} column ) - ( x_1 \times (n-1)^{th} column ) ]
                                                                    x_k \quad \cdots \quad x_k^{k-1}
                                                           x_2^{k-2}(x_2-x_1)
                                                          x_2^{k-1}(x_2-x_1)
                                                                                                                                   x_2^{k-2}(x_2-x_1) x_2^{k-1}(x_2-x_1)
                                                                                                        \begin{bmatrix} \vdots & \vdots \\ 1 & x_k - x_1 \end{bmatrix}
                                                             x_k^{k-1}(x_k-x_1)
                                                                                                        1 \quad x_{k+1} - x_1 \quad \cdots \quad x_{k+1}^{k-1}(x_{k+1} - x_1) \quad x_{k+1}^{k-1}(x_{k+1} - x_1)
                     x_{k+1}^{k-1}(x_{k+1}-x_1) x_{k+1}^{k-1}(x_{k+1}-x_1)
                                                                                                 [ : cofactor expansion for first row ]
                                                            x_k^{k-1}(x_k-x_1)
= \prod_{i=2}^{k+1} (x_i - x_1) \begin{bmatrix} 1 & \cdots & x_2^{k-2} & x_2^{k-1} \\ \vdots & \ddots & \vdots & \vdots \\ 1 & \cdots & x_k^{k-2} & x_k^{k-1} \end{bmatrix}
                                                                 =\prod_{i=2}^{k+1}(x_i-x_1)\times\prod_{j=2}^{k}\prod_{i=j+1}^{k+1}(x_i-x_j) [ : P(k)\equiv T ]
                            1 \quad \cdots \quad x_{k+1}^{k-1} \quad x_{k+1}^{k-1}
= \prod_{j=1}^{k} \prod_{i=j+1}^{k+1} (x_i - x_j), so that P(k+1) \equiv T
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What is the condition that makes the determinant of A non-zero?

 $\det A = \prod_{i=1}^{n-1} \prod_{i=i+1}^n (x_i - x_i)$ is non-zero when there is no a_i b such that $a \neq b$ and $x_a = x_b$

 $\div \, \det A = \prod_{j=1}^{n-1} \prod_{i=j+1}^n (x_i - x_j)$ (for $n \geq 2$) [for $n=1, \, \det A = 1$]

Assume that the determinant of A is non-zero, then, what is the solution of linear equation, Aw = y, with respect to w? 'the determinant of A is non-zero' means that A is invertible. so, we can get solution to compute $w = A^{-1}y$

Suppose that n > d + 1. Then, we cannot comput the inverse of A since A is not a square matrix. In this case, how can we solve the linear equation Aw = y?

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with SVD, we can write A = U\Sigma V^T [ U : n \times n orthogonal matrix, V : (d+1) \times (d+1) orthogonal matrix, \Sigma : n \times (d+1) matrix ] [ Let rank A = r, then \Sigma = \begin{bmatrix} \Sigma_r & 0 \\ 0 & 0 \end{bmatrix}, \Sigma_r : r \times r diagonal matrix ] notice that U^T = U^{-1} and V^T = V^{-1}, and pseudo inverse matrix \Sigma^* = \begin{bmatrix} \Sigma_r^{-1} & 0 \\ 0 & 0 \end{bmatrix} [ : \Sigma \Sigma^* \Sigma = \begin{bmatrix} \Sigma_r & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \Sigma_r^{-1} & 0 \\ 0 & 0 \end{bmatrix} \Sigma = \Sigma and \Sigma^* \Sigma \Sigma^* = \begin{bmatrix} \Sigma_r^{-1} & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \Sigma_r & 0 \\ 0 & 0 \end{bmatrix} \Sigma = \Sigma and \Sigma^* \Sigma \Sigma^* = \begin{bmatrix} \Sigma_r^{-1} & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \Sigma_r & 0 \\ 0 & 0 \end{bmatrix} \Sigma^* = \Sigma ] then, A^+ = V\Sigma^+ U^T [ : A^+ AA^+ = VE^+ U^T UEV^T VE^+ U^T = VE^+ EE^+ U^T = VE^+ U^T = A^+, and AA^+ A = U\Sigma V^T V\Sigma^+ U^T U\Sigma V^T = U\Sigma \Sigma^+ \Sigma V^T = U\Sigma V^T = A] on the other hand, A^+ A = V\Sigma^+ U^T U\Sigma V^T = V\Sigma^+ \Sigma V^T = V \begin{bmatrix} I_r & 0 \\ 0 & 0 \end{bmatrix} V^T \neq I_{d+1}. but, if rank A = d + 1, then we can write \Sigma = \begin{bmatrix} \Sigma_r \\ 0 \end{bmatrix} so that \Sigma^+ \Sigma = \begin{bmatrix} \Sigma_r^{-1} & 0 \end{bmatrix} \begin{bmatrix} \Sigma_r \\ 0 \end{bmatrix} = I_{d+1} so, if rank A = d + 1, that is, if A has linearly independent columns, than A^+ A = I so that Aw = y \Rightarrow A^+ Aw = A^+ y \Rightarrow w = A^+ y.

on the other hand, if A has linearly independent columns, A^T A is invertible [ : Col A = (Nul A^T)^\perp so that A^T Ax = 0 \Rightarrow Ax = 0 \Rightarrow x = 0 ] so, A^+ is existed by (A^T A)^{-1} A^T [ : A^+ A = (A^T A)^{-1} A^T A = I ] : w we can solve the linear equation Aw = y by computing w = A^+ y = (A^T A)^{-1} A^T y
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