# On the Inversion of Vandermonde Matrices

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#### Abstract

A novel and simple recursive algorithm for inverting Vandermonde matrix and its generalized form is presented. The algorithm is suitable for classroom use in both numerical as well as symbolic computation.

# 1 Introduction

The importance of the Vandermonde matrix is well known. Inversion of this matrix is necessary in many areas of applications such as polynomial interpolation [4, 10], digital signal processing [2], and control theory [6], to mention a few. See also for example Klinger [8], Kalman [7]. However, an explicit recursive formula for the inversion of Vandermonde matrices seems unavailable in most linear algebra textbooks.

The purpose of this paper is to present a novel and simple recursive algorithm for inverting Vandermonde matrix, as well as its generalized (or confluent) form, in a way more readily accessible for use in classroom and suitable for both numerical as well as symbolic computation.

# 2 Preliminaries and notations

Let *m* be a nonnegative integer. For the sequence  $1, (s - \lambda), \ldots, (s - \lambda)^{m-1}$  of polynomials we write  $\mathbf{s}(\lambda, m) = [1, (s - \lambda), \ldots, (s - \lambda)^{m-1}]^T$ . In particular,

$$\mathbf{s}(0,m) = [1, s, \dots, s^{m-1}]^T$$

Let  $\lambda_1, \lambda_2, \ldots, \lambda_r$  be given distinct zeros of the polynomial

$$p(s) = (s - \lambda_1)^{n_1} \cdots (s - \lambda_r)^{n_r}$$

with  $n_1 + \ldots + n_r = n$ . The generalized (or confluent) Vandermonde matrix related to the zeros of p(s) is known to be

$$V = \left[V_1 V_2 \cdots V_r\right],\tag{1}$$

where the block matrix  $V_k = V(\lambda_k, n_k)$  is of order  $n \times n_k$ , having elements  $V(\lambda_k, n_k)_{ij} = {i-1 \choose j-1} \lambda_k^{i-j}$  for  $i \ge j$  and zero otherwise  $(k = 1, 2, ..., r; i = 1, 2, ..., n; j = 1, 2, ..., n_k)$ . More specifically,  $V_k$  is the  $n \times n_k$  matrix of coefficients that appears in the truncated Taylor expansion at  $\lambda_k$ , modulo  $(s - \lambda_k)^{n_k}$ , of  $\mathbf{s}(0, n)$ . That is,

$$\mathbf{s}(0,n) = V(\lambda_k, n_k)\mathbf{s}(\lambda_k, n_k) \mod (s - \lambda_k)^{n_k}.$$

In the case the zeros  $\lambda_1, \ldots, \lambda_r$  of p(s) are simple, we have the usual Vandermonde matrix, namely,

$$V = \begin{bmatrix} 1 & 1 & \cdots & 1\\ \lambda_1 & \lambda_2 & \cdots & \lambda_r\\ \vdots & \vdots & & \vdots\\ \lambda_1^{n_1-1} & \lambda_2^{n_2-1} & \cdots & \lambda_r^{n_r-1} \end{bmatrix}.$$

It will be shown that the inverse of the generalized Vandermonde matrix V in (1) has a form

$$V^{-1} = \begin{bmatrix} W_1 \\ W_2 \\ \vdots \\ W_r \end{bmatrix},$$

where each block matrix  $W_k$  is of order  $n_k \times n$ , and may be computed by means of a recursive procedure.

Generally, the inverse of the usual Vandermonde matrix [3], as well as the inverse of the generalized Vandermonde matrix [9] are based on using interpolation polynomials.

Our approach is based on using the Leverrier-Faddeev algorithm [1, 5, 10], which states that the resolvent of a given  $n \times n$  matrix A is given by

$$(sI - A)^{-1} = \frac{B_1 s^{n-1} + B_2 s^{n-2} + \dots + B_n}{s^n + a_1 s^{n-1} + \dots + a_n},$$
(2)

where  $det(sI - A) = s^n + a_1 s^{n-1} + \dots + a_n$  is the characteristic polynomial

of the matrix A, and all the  $B_j$  matrices are of order  $n \times n$ , satisfying

$$B_{1} = I, \qquad a_{1} = -\frac{1}{1} \operatorname{tr}(AB_{1}), \\ B_{2} = AB_{1} + a_{1}I, \qquad a_{2} = -\frac{1}{2} \operatorname{tr}(AB_{2}), \\ \vdots \qquad \vdots \\ B_{n} = AB_{n-1} + a_{n-1}I, \qquad a_{n} = -\frac{1}{n} \operatorname{tr}(AB_{n})$$
(3)

with  $0 = AB_n + a_n I$  terminating as a check of computation. Here tr stands for the trace of a matrix.

#### 3 Main result

Let  $J = \text{diag}(J_1, \ldots, J_r)$  be the block diagonal matrix, where

$$J_{k} = J(\lambda_{k}, n_{k}) = \begin{bmatrix} \lambda_{k} & 1 & 0 & \cdots & 0 \\ 0 & \lambda_{k} & 1 & & \vdots \\ 0 & & \ddots & \ddots & 0 \\ \vdots & & \lambda_{k} & 1 \\ 0 & \cdots & 0 & 0 & \lambda_{k} \end{bmatrix}$$

is the  $n_k \times n_k$  Jordan block with eigenvalue  $\lambda_k$ . Then J has characteristic polynomial det $(sI - J) = (s - \lambda_1)^{n_1} \cdots (s - \lambda_r)^{n_r} = p(s)$ .

Substituting A = J in equations (2) and (3) of the Leverrier-Faddeev algorithm, we see immediately that

$$p(s)(sI - J)^{-1} = B_1 s^{n-1} + B_2 s^{n-2} + \dots + B_n,$$
(4)

where

$$B_{1} = I, B_{2} = JB_{1} + a_{1}I, \dots B_{n} = JB_{n-1} + a_{n-1}I, 0 = JB_{n} + a_{n}I.$$

 $J = \operatorname{diag}(J_1, \ldots, J_r)$  being block diagonal, so are all the  $B_j$  matrices. In fact,

$$B_j = \text{diag}(B_{j,1}, B_{j,2}, \dots, B_{j,r}), \quad j = 1, 2, \dots, n$$

and each block matrix  $B_{j,k}$  is of order  $n_k \times n_k$ , satisfying

$$\begin{array}{rcl}
B_{1,k} &=& I_k, \\
B_{2,k} &=& J_k B_{1,k} + a_1 I_k, \\
& \cdots & & \\
B_{n,k} &=& J_k B_{n-1,k} + a_{n-1} I_k, \\
0 &=& J_k B_{n,k} + a_n I_k,
\end{array}$$
(5)

where  $I_k$  is the  $n_k \times n_k$  identity matrix. Let us now put

$$p_k(s) = \frac{p(s)}{(s - \lambda_k)^{n_k}}, \qquad k = 1, \dots, r.$$

Define also the  $n_k$ -dimensional column vector  $\theta_k = [0, \dots, 0, 1]^T$ , and write  $\theta = [\theta_1^T, \dots, \theta_r^T]^T$ .

If we postmultiply both sides of equation (4) by the column vector  $\theta$ , we easily get

$$\begin{bmatrix} p_1(s)\mathbf{s}(\lambda_1, n_1) \\ \vdots \\ p_r(s)\mathbf{s}(\lambda_r, n_r) \end{bmatrix} = \begin{bmatrix} H_1 \\ \vdots \\ H_r \end{bmatrix} \mathbf{s}(0, n).$$
(6)

Each  $H_k$  is of the form

$$H_k = \left[ \begin{array}{ccc} B_{n,k}\theta_k & \cdots & B_{1,k}\theta_k \end{array} \right]$$
(7)

and has order  $n_k \times n$ .

Comparing in turn for k = 1, 2, ..., r the truncated Taylor expansions at  $\lambda_k$ , modulo  $(s - \lambda_k)^{n_k}$ , of both sides in (6) and putting these results together, we get

diag
$$(P_1,\ldots,P_r) = \begin{bmatrix} H_1 \\ \vdots \\ H_r \end{bmatrix} \begin{bmatrix} V_1 & \cdots & V_r \end{bmatrix},$$

where each block  $P_k$  is a  $n_k \times n_k$  upper triangular matrix given by

$$P_{k} = p_{k}(J_{k}) = \sum_{j=0}^{n_{k}-1} \frac{p_{k}^{(j)}(\lambda_{k})}{j!} (N_{k})^{j}.$$

It is noted here that  $N_k = J(0, n_k) = J_k - \lambda_k I_k$  is nilpotent of order  $n_k$ . If we can show that each  $P_k$  is invertible, then

$$V^{-1} = \begin{bmatrix} P_1^{-1}H_1 \\ \vdots \\ P_r^{-1}H_r \end{bmatrix}.$$
 (8)

To this end we require the following lemma which is an easy consequence of the partial fraction expansion of 1/p(s) and the fact that  $N_k$  is nilpotent.

Lemma 1 Let there be given the partial fraction expansion

$$\frac{1}{p(s)} = \sum_{k=1}^{r} \left( \frac{K_{k,n_k}}{(s-\lambda_k)^{n_k}} + \frac{K_{k,n_k-1}}{(s-\lambda_k)^{n_k-1}} + \dots + \frac{K_{k,1}}{s-\lambda_k} \right).$$

*Then for* k = 1, 2, ..., r

$$P_k^{-1} = \sum_{j=0}^{n_k-1} K_{k,j}(N_k)^j = \mathcal{K}_k(J_k),$$

where the polynomial  $\mathcal{K}_k(s)$  is given by

$$\mathcal{K}_k(s) = K_{k,n_k} + K_{k,n_k-1}(s-\lambda_k) + \dots + K_{k,1}(s-\lambda_k)^{n_k-1}.$$

Putting the above results together with equations (7) and (8), we are now ready to state our main result:

**Theorem 1** The inverse of  $V = [V_1V_2...V_r]$  related to the distinct zeros  $\lambda_1, \ldots, \lambda_r$  of p(s) is given by

$$V^{-1} = \begin{bmatrix} W_1 \\ W_2 \\ \vdots \\ W_r \end{bmatrix},$$

where each block matrix

$$W_k = W(\lambda_k, n_k) = \left[ \begin{array}{ccc} \mathcal{K}_k(J_k) B_{n,k} \theta_k & \mathcal{K}_k(J_k) B_{n-1,k} \theta_k & \cdots & \mathcal{K}_k(J_k) B_{1,k} \theta_k \end{array} \right]$$

is of order  $n_k \times n$ .

Taking into account of (5), we find that  $\mathcal{K}_k(J_k)B_{j,k} = B_{j,k}\mathcal{K}_k(J_k), j = 1, 2, \ldots, n$ , so that  $B_{1,k}\mathcal{K}_k(J_k) = \mathcal{K}_k(J_k)$ , and for  $j = 2, \ldots, n$ 

$$B_{j,k}\mathcal{K}_k(J_k) = J_k B_{j-1,k}\mathcal{K}_k(J_k) + a_{j-1}\mathcal{K}_k(J_k)$$
  
=  $(\lambda_k I_k + N_k)B_{j-1,k}\mathcal{K}_k(J_k) + a_{j-1}\mathcal{K}_k(J_k).$ 

Moreover,  $(\lambda_k I_k + N_k)B_{n,k}\mathcal{K}_k(J_k) + a_n\mathcal{K}_k(J_k) = 0.$ 

# 4 The Algorithm

Based on the results obtained in the last section, we are now ready to give a recursive algorithm for inverting generalized Vandermonde matrix.

The Algorithm:

Let  $\lambda_1, \lambda_2, \ldots, \lambda_r$  be distinct zeros of the polynomial

$$p(s) = (s - \lambda_1)^{n_1} \cdots (s - \lambda_r)^{n_r}$$
$$= s^n + a_1 s^{n-1} + \cdots + a_n$$

given together with the partial fraction expansion of

$$\frac{1}{p(s)} = \sum_{k=1}^{r} \left( \frac{K_{k,n_k}}{(s-\lambda_k)^{n_k}} + \frac{K_{k,n_k-1}}{(s-\lambda_k)^{n_k-1}} + \dots + \frac{K_{k,1}}{s-\lambda_k} \right).$$

For each  $k \in \{1, 2, ..., r\}$ , compute recursively polynomials  $h_1, h_2, ..., h_n$  of degree at most  $n_k - 1$  by means of the following scheme:

$$h_1(s) = K_{k,n_k} + sK_{k,n_{k-1}} + \dots + s^{n_k - 1}K_{k,1},$$
  

$$h_2(s) = (\lambda_k + s)h_1(s) + a_1h_1(s) \mod s^{n_k},$$
  

$$h_3(s) = (\lambda_k + s)h_2(s) + a_2h_1(s) \mod s^{n_k},$$
  

$$\vdots$$
  

$$h_n(s) = (\lambda_k + s)h_{n-1}(s) + a_{n-1}h_1(s) \mod s^{n_k},$$

terminating at

$$0 = (\lambda_k + s)h_n(s) + a_nh_1(s) \mod s^{n_k}.$$

Obtain a block matrix  $W_k = W(\lambda_k, n_k)$  of order  $n_k \times n$  via the equality

$$\left[\begin{array}{cccc} s^{n_k-1} & s^{n_k-2} & \cdots & 1\end{array}\right] W(\lambda_k, n_k) = \left[\begin{array}{cccc} h_n & h_{n-1} & \cdots & h_1\end{array}\right].$$

The inverse of the generalized Vandermonde matrix V related to the distinct zeros  $\lambda_1, \lambda_2, \ldots, \lambda_r$  of p(s) may then be given by

$$V^{-1} = [V(\lambda_1, n_1)V(\lambda_2, n_2) \cdots V(\lambda_r, n_r)]^{-1}$$
$$= \begin{bmatrix} W(\lambda_1, n_1) \\ W(\lambda_2, n_2) \\ \vdots \\ W(\lambda_r, n_r) \end{bmatrix}.$$

Let us now give some supplementary remarks on the above algorithm.

- (i) A check on the accuracy of the computation of polynomials  $h_1, \ldots, h_n$  is provided by the last polynomial  $(\lambda_k + s)h_n(s) + a_nh_1(s)$ , which should result identically in the zero polynomial 0 when modulo  $s^{n_k}$  is performed.
- (ii) The coefficients  $a_1, a_2, \ldots, a_n$  of the polynomial p(s) may be recursively computed using (3) with  $A = \text{diag}(\underbrace{\lambda_1, \ldots, \lambda_1}_{n_1}, \ldots, \underbrace{\lambda_r, \ldots, \lambda_r}_{n_r})$ .
- (iii) The partial fraction coefficients  $K_{k,n_k}, K_{k,n_k-1}, \ldots, K_{k,1}$  used in the construction of the starting polynomial  $h_1(s)$  may be obtained by expanding

$$\frac{1}{p_k(s)} = \sum_{j=0}^{n_k-1} K_{k,n_k-j}(s-\lambda_k)^j + \cdots$$

in powers of  $(s - \lambda_k)$ . They may also be recursively computed using the following scheme:

$$K_{k,n_k} = 1/p_{k,0} ,$$
  

$$K_{k,n_k-j} = -\frac{\sum_{i=1}^{j} p_{k,i} K_{k,n_k-j+i}}{p_{k,0}} \quad (j = 1, \dots, n_k - 1),$$

where  $p_k(s) = \sum_{j=0}^{n-n_k} p_{k,j}(s-\lambda_k)^j$ .

# 5 Illustrative example

The following example will serve to illustrate the recursive algorithm presented above. Let the generalized Vandermonde matrix V in (1) be given by

$$V = \begin{bmatrix} 1 & 0 & 0 & 1 \\ \lambda_1 & 1 & 0 & \lambda_2 \\ \lambda_1^2 & 2\lambda_1 & 1 & \lambda_2^2 \\ \lambda_1^3 & 3\lambda_1^2 & 3\lambda_1 & \lambda_2^3 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 1 \\ -2 & 1 & 0 & 3 \\ 4 & -4 & 1 & 9 \\ -8 & 12 & -6 & 27 \end{bmatrix},$$

for which  $\lambda_1 = -2, n_1 = 3$ , and  $\lambda_2 = 3, n_2 = 1$ . The coefficients of the polynomial  $p(s) = (s+2)^3(s-3)$  are given by  $a_1 = 3, a_2 = -6, a_3 = -28$ ,

and  $a_4 = -24$ . It is easy to determine the partial fraction expansion of 1/p(s) to be

$$\frac{1}{(s+2)^3(s-3)} = \frac{-\frac{1}{5}}{(s+2)^3} + \frac{-\frac{1}{25}}{(s+2)^2} + \frac{-\frac{1}{125}}{s+2} + \frac{\frac{1}{125}}{s-3}$$

Let us consider first the case  $\lambda_1 = -2$ . Clearly

$$h_1(s) = -\frac{1}{5} - \frac{s}{25} - \frac{s^2}{125}.$$

Then

$$h_2(s) = (-2+s)h_1(s) + 3h_1(s) \mod s^3$$
  
=  $-\frac{1}{5} - \frac{6s}{25} - \frac{6s^2}{125}$ ,  
$$h_3(s) = (-2+s)h_2(s) - 6h_1(s) \mod s^3$$
  
=  $\frac{8}{5} + \frac{13s}{25} - \frac{12s^2}{125}$ ,  
$$h_4(s) = (-2+s)h_3(s) - 28h_1(s) \mod s^3$$
  
=  $\frac{12}{5} + \frac{42s}{25} + \frac{117s^2}{125}$ .

As a check of computation, we verify that

$$(-2+s)h_4(s) - 24h_1(s) \mod s^3 = \frac{117s^3}{125} \mod s^3 = 0.$$

Thus it follows from  $\begin{bmatrix} s^2 & s & 1 \end{bmatrix} W_1 = \begin{bmatrix} h_4 & h_3 & h_2 & h_1 \end{bmatrix}$  that

$$W_1 = \begin{bmatrix} \frac{117}{125} & -\frac{12}{125} & -\frac{6}{125} & -\frac{1}{125} \\ \frac{42}{25} & \frac{13}{25} & -\frac{6}{25} & -\frac{1}{25} \\ \frac{12}{5} & \frac{8}{5} & -\frac{1}{5} & -\frac{1}{5} \end{bmatrix}$$

•

Similarly, for  $\lambda_2 = 3$  we find that

$$h_1(s) = \frac{1}{125},$$
  

$$h_2(s) = (3+s)h_1(s) + 3h_1(s) \mod s = \frac{6}{125},$$
  

$$h_3(s) = (3+s)h_2(s) - 6h_1(s) \mod s = \frac{12}{125},$$
  

$$h_4(s) = (3+s)h_3(s) - 28h_1(s) \mod s = \frac{8}{125}.$$

and

$$(3+s)h_4(s) - 24h_1(s) \mod s = 3 \cdot \frac{8}{125} - \frac{24}{125} = 0$$
  
Then  $\begin{bmatrix} 1 \end{bmatrix} W_2 = \begin{bmatrix} h_4 & h_3 & h_2 & h_1 \end{bmatrix}$  gives  
$$W_2 = \begin{bmatrix} \frac{8}{125} & \frac{12}{125} & \frac{6}{125} & \frac{1}{125} \end{bmatrix}.$$

Finally, we have

$$V^{-1} = \begin{bmatrix} W_1 \\ W_2 \end{bmatrix} = \begin{bmatrix} \frac{117}{125} & -\frac{12}{125} & -\frac{6}{125} & -\frac{1}{125} \\ \frac{42}{25} & \frac{13}{25} & -\frac{6}{25} & -\frac{1}{25} \\ \frac{12}{5} & \frac{8}{5} & -\frac{1}{5} & -\frac{1}{5} \\ \frac{8}{125} & \frac{12}{125} & \frac{6}{125} & \frac{1}{125} \end{bmatrix}.$$

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