

# Super linear algebra

## Block matrices

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**Abstract** - This seminar has three parts . Part one introduce the notion of super matrices . Part two discuss the notions of super linear algebra , the characteristic super polynomial of a linear transformation , the minimal polynomial , and Cayley–Hamilton theorem . In part three we have an application of using super matrices for solving BVP .

**Keywords** :*Super matrices ;Super linear algebra ; Block matrices .*

### PART ONE

#### INTRODUCTION

In this part we shall explain the notion of super matrices , super vector space and give some examples on it .

\* Super matrices

Definition(1.1):A super matrix is a matrix whose entries are matrices .

Example(1.1): Let 
$$A = \begin{bmatrix} 0 & 1 & 7 & 2 & 5 \\ 3 & 5 & 16 & 2 & 4 \\ \hline 6 & 8 & 0 & 5 & 3 \\ 9 & 9 & 7 & 2 & 3 \\ 7 & 0 & 5 & 2 & 4 \end{bmatrix}$$

Then A is a super matrix with two rows and two columns .

Now let

$$A_{11} = \begin{bmatrix} 0 & 1 \\ 3 & 5 \end{bmatrix}, A_{12} = \begin{bmatrix} 7 & 2 & 5 \\ 16 & 2 & 4 \end{bmatrix}, A_{21} = \begin{bmatrix} 6 & 8 \\ 9 & 9 \\ 7 & 0 \end{bmatrix}, A_{22} = \begin{bmatrix} 0 & 5 & 3 \\ 7 & 2 & 3 \\ 5 & 2 & 4 \end{bmatrix}$$

Then we can write

$$A = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}.$$

Note :Sometimes a super matrix is called a block matrix .And we call the entries sub matrices or blocks .

Definition(1.2):A super matrix of the form

$$D = \begin{bmatrix} D_{11} & 0 & 0 \\ 0 & D_{22} & 0 \\ \hline 0 & 0 & D_{nn} \end{bmatrix} \text{ is called a}$$

super diagonal matrix .

Example(1.2): The super matrix

$$A = \begin{bmatrix} 5 & 6 & 0 & 0 & 0 & 0 & 0 & 0 \\ 8 & 4 & 0 & 0 & 0 & 0 & 0 & 0 \\ \hline 0 & 0 & 7 & 6 & 5 & 0 & 0 & 0 \\ 0 & 0 & 4 & 3 & 2 & 0 & 0 & 0 \\ \hline 0 & 0 & 0 & 0 & 0 & 1 & 2 & 5 \end{bmatrix} \text{ is a}$$

super diagonal matrix .

Definition(1.3):A super matrix D where  $D_{ii}$  are square matrices is called a symmetrically partitioned super matrix .

Example(1.3): The super matrix

$$A = \left[ \begin{array}{cc|cc|c} 5 & 4 & 1 & 3 & 4 & 0 \\ 3 & 2 & 6 & 8 & 8 & 8 \\ \hline 5 & 6 & 6 & 8 & 0 & 6 \\ 3 & 8 & 5 & 8 & 1 & 5 \\ \hline 4 & 4 & 4 & 8 & 0 & 4 \\ 5 & 0 & 1 & 7 & 9 & 8 \end{array} \right] \text{ is a}$$

symmetrically partitioned super matrix .

## PART TWO

### SUPER LINEAR ALGEBRA

In this part we shall define the notion of super linear algebra and give some examples on it; also we shall discuss the characteristic super polynomial of a super matrix and Cayley-Hamilton theorem .

Definition(2.1) : Let  $V$  be a super vector space over the field  $F$ . We say  $V$  is a super linear algebra over  $F$  if and only if for every pair of super vectors  $\alpha, \beta$  the product of  $\alpha$  and  $\beta$  denoted by  $\alpha\beta$  is defined in  $V$  in such a way that :

- $\alpha(\beta\gamma) = (\alpha\beta)\gamma \quad \forall \alpha, \beta, \gamma \in V$  .
- $(\alpha + \beta)\gamma = \alpha\gamma + \beta\gamma$  and  $\gamma(\alpha + \beta) = \gamma\alpha + \gamma\beta \quad \forall \alpha, \beta, \gamma \in V$  .
- $c(\alpha\beta) = (c\alpha)\beta = \alpha(c\beta) \quad \forall \alpha, \beta \in V, c \in F$  .

Definition (2.2) : If a super linear algebra  $V$  has an element  $I$  such that  $Ia = aI, \forall a \in V$  , then we call  $V$  a super linear algebra with identity .

Definition (2.3) : A super linear algebra  $V$  is called commutative if  $ab = ba \quad \forall a, b \in V$  .

Example (2.1): Let

$$V = \{ [x_1 \ x_2 \mid x_3 \ x_4 \mid x_5], x_i \in Q, 1 \leq i \leq 5 \}$$

be a super vector space over  $R$ . Define for

$$\alpha, \beta \in V ;$$

$$\alpha = [x_1 \ x_2 \mid x_3 \ x_4 \mid x_5]$$

$$\text{and } \beta = [y_1 \ y_2 \mid y_3 \ y_4 \mid y_5]$$

$$\alpha\beta = [x_1y_1 \ x_2y_2 \mid x_3y_3 \ x_4y_4 \mid x_5y_5]$$

Where  $x_i, y_i \in R, 1 \leq i \leq 5$ .

Then  $V$  is a super linear algebra over  $R$  , and it's a commutative super linear algebra with unity  $[1 \ 1 \mid 1 \ 1 \mid 1]$ .

Example (2.2) : Let

$$V = \left\{ \left[ \begin{array}{cc|c} x_1 & x_2 & x_3 \\ x_4 & x_5 & x_6 \\ \hline x_7 & x_8 & x_9 \end{array} \right] : x_i \in Q, 1 \leq i \leq 9 \right\}$$

then  $V$  is a super linear algebra over  $Q$  with the usual multiplication of matrices , and it's clear that  $V$  is a non-commutative super linear

$$\text{algebra with unity } \left[ \begin{array}{cc|c} 1 & 0 & 0 \\ 0 & 1 & 0 \\ \hline 0 & 0 & 1 \end{array} \right] .$$

Example (2.3): Let  $V$  be a super vector space over  $F$  then  $SL(V, V)$  is a super linear algebra over  $F$  , under the operation of functions composition .

Definition (2.3) : Let

$$V = \left\{ [A_1 \ A_2 \ \cdots \ A_k] : A_i \text{ are row vectors with entries from a field } F \right\}$$

and let  $T \in SL(V, V)$  ,  $T = [T_1 \ T_2 \ \cdots \ T_k]$ .

The k-tuple  $c = (c_1, c_2, \dots, c_k)$  is called a

characteristic super value of  $T$  if there is a non-zero super vector  $a \in V$  such that :

$$\begin{aligned} T(a) &= ca \\ &= (c_1, c_2, \dots, c_k) [a_1 \ a_2 \ \dots \ a_k] \\ &= [c_1 a_1 \ c_2 a_2 \ \dots \ c_k a_k] \end{aligned}$$

And we call  $a$  a characteristic super vector of  $T$  associated with  $c = (c_1, c_2, \dots, c_k)$ .

**Definition (2.4) :** Let  $A$  be a square super diagonal matrix whose diagonal matrices are also squares, then the super determinant of  $A$  is defined as :

$$|A| = \begin{bmatrix} |A_1| & \cdots & \cdots \\ \vdots & |A_2| & \cdots \\ \vdots & \vdots & \ddots \\ & & & |A_n| \end{bmatrix} = [|A_1| \ |A_2| \ \cdots \ |A_n|]$$

where  $|A_i|$  is the determinant of  $A_i$ .

$i = 1, 2, \dots, n$ .

Note : if  $A$  is a square super diagonal matrix whose diagonal matrices are also squares, then the super determinant of  $A$  is a super row vector.

**Example (2.4) :** Find the determinant of  $A$  where :

$$A = \left[ \begin{array}{cc|cc|cc|c} 7 & 3 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 0 & 0 \\ \hline 0 & 0 & 5 & 8 & 0 & 0 & 0 \\ 0 & 0 & 4 & 6 & 0 & 0 & 0 \\ \hline 0 & 0 & 0 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 & 2 & 2 & 0 \\ \hline 0 & 0 & 0 & 0 & 0 & 0 & 7 \end{array} \right]$$

Solution :

$$\begin{aligned} |A| &= \left[ \begin{array}{cc|cc|c} 7 & 3 & 5 & 8 & 1 & 1 \\ 1 & 1 & 4 & 6 & 2 & 2 \\ \hline 4 & -2 & 0 & 7 \end{array} \right] \\ &= [4 \ | \ -2 \ | \ 0 \ | \ 7] \end{aligned}$$

**Definition (2.5) :** A super polynomial is a polynomial of the form :

$$P(X) = [P_1(X) \ P_2(X) \ \dots \ P_n(X)];$$

$p_i(x)$  is a polynomial  $1 \leq i \leq n$ .

**Example (2.5) :** Let

$$V = \{[R^2(X) \ R^7(X) \ R^1(X)]\}$$

$R^i(X)$  is a polynomial of degree less than or equal  $i$  with coefficients from the real numbers, then  $V$  is a super vector space over  $R$  with the operation of addition and scalar multiplication, ( the dimension of  $V$  is  $3+8+2=13$  ). It's clear that  $V$  is a super linear algebra over  $R$  if we define

for  $P_1(X), P_2(X) \in P(X)$  and

$$P_1(X) = [P_1^2(X) \ P_1^7(X) \ P_1^1(X)]$$

$$P_2(X) = [P_2^2(X) \ P_2^7(X) \ P_2^1(X)]$$

$$P_1(X)P_2(X) = [P_1^2(X)P_2^2(X) \ P_1^7(X)P_2^7(X) \ P_1^1(X)P_2^1(X)]$$

**Definition (2.6) :** Let

$$A = \begin{bmatrix} A_1 & 0 & \cdots & 0 \\ 0 & A_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & A_n \end{bmatrix}$$

be a square super diagonal square matrix with entries from a field  $F$  where  $A_i$  is a square matrix  $i = 1, 2, \dots, n$ . Then the super polynomial :

$$P(X) = \det(xI - A)$$

$$\begin{aligned} &= [\det(xI - A_1) \ \det(xI - A_2) \ \cdots \ \det(xI - A_n)] \\ &= [f_1 \ f_2 \ \cdots \ f_n] \end{aligned}$$

is called the characteristic super polynomial (Super characteristic polynomial) of  $A$ .

Note : The order of the super characteristic polynomial  $P(X)$  in the previous definition is  $(k_1, k_2, \dots, k_n)$  where  $k_i$  is the order of the matrix  $A_i, 1 \leq i \leq n$ .

Example (2.6) : Let

$$A = \left[ \begin{array}{cc|cc} 2 & 1 & 0 & 0 \\ 3 & 1 & 0 & 0 \\ \hline 0 & 0 & 4 & 2 \\ 0 & 0 & 2 & 4 \end{array} \right], \text{ then the characteristic}$$

super polynomial of A is

$$|xI - A| = \left[ \begin{array}{cc|cc} x-2 & 1 & 0 & 0 \\ 3 & x-1 & 0 & 0 \\ \hline 0 & 0 & x-4 & 2 \\ 0 & 0 & -2 & x-4 \end{array} \right]$$

$$= [(x-2)(x-1)-3 \mid (x-4)^2-4]$$

$$= [x^2 - 3x - 1 \mid x^2 - 8x - 12]$$

To find the characteristic super value we must find the zeros of the characteristic super polynomial :

$$P(x) = [P_1(x) \quad P_2(x)]$$

$$P_1(x) = x^2 - 3x - 1$$

$$P_2(x) = x^2 - 8x - 12$$

$$\text{If } P_1(x) = x^2 - 3x - 1 = 0 \text{ then}$$

$$x = \frac{3 \pm \sqrt{9+4}}{2} = \frac{3 \pm \sqrt{13}}{2}$$

$$\text{And if } P_2(x) = x^2 - 8x - 12 = 0 \text{ then}$$

$$x = \frac{8 \pm \sqrt{64+48}}{2} = \frac{8 \pm \sqrt{112}}{2} = 4 \pm 2\sqrt{7}$$

So the characteristic super values are :

$$C_1 = \left[ \frac{3 + \sqrt{13}}{2} \mid 4 + 2\sqrt{7} \right]$$

$$C_2 = \left[ \frac{3 + \sqrt{13}}{2} \mid 4 - 2\sqrt{7} \right]$$

$$C_3 = \left[ \frac{3 - \sqrt{13}}{2} \mid 4 + 2\sqrt{7} \right]$$

$$C_4 = \left[ \frac{3 - \sqrt{13}}{2} \mid 4 - 2\sqrt{7} \right]$$

$$\text{Example(2.7): Let } A = \left[ \begin{array}{cc|cc} 2 & 1 & 0 & 0 \\ 1 & 2 & 0 & 0 \\ \hline 0 & 0 & 1 & -1 \\ 0 & 0 & 1 & 1 \end{array} \right]$$

Then the super characteristic polynomial of A is :

$$P(\lambda) = [\lambda^2 - 4\lambda + 3 \mid \lambda^2 - 2\lambda + 2].$$

And if we solve  $P(\lambda) = 0$  then the super eigenvalues are :

$$\lambda_1 = [1 \mid 1+i]$$

$$\lambda_2 = [3 \mid 1+i]$$

$$\lambda_3 = [1 \mid 1-i]$$

$$\lambda_4 = [3 \mid 1-i]$$

With the associated super eigenvectors :

$$V_1 = \begin{bmatrix} -1 \\ 1 \\ i \\ 1 \end{bmatrix} \quad V_2 = \begin{bmatrix} 1 \\ 1 \\ i \\ 1 \end{bmatrix} \quad V_3 = \begin{bmatrix} -1 \\ 1 \\ -i \\ 1 \end{bmatrix} \quad V_4 = \begin{bmatrix} 1 \\ 1 \\ -i \\ 1 \end{bmatrix}$$

Related to the super eigenvectors respectively .

Definition (2.7) : Let  $T = [T_1 \quad T_2 \quad \dots \quad T_n]$

be a linear operator on an infinite dimensional

super vector space  $V = [V_1 \ V_2 \ \cdots \ V_n]$  of dimension  $(n_1, n_2, \dots, n_n)$  over the field  $F$ , the minimal super polynomial for  $T$  is the unique monic super generator of the super ideal of super polynomials over  $F$  which super annihilate  $T$ .

This definition means that the minimal super polynomial  $P = [P_1(x) \ P_2(x) \ \cdots \ P_n(x)]$  for the linear operator  $T$  is uniquely determined by the following properties :

- 1)  $P = [P_1 \ P_2 \ \cdots \ P_n]$  is a monic polynomial over the field  $F$ .
- 2)  $P(T) = [P_1(T_1) \ P_2(T_2) \ \cdots \ P_n(T_n)] = [0 \ 0 \ \cdots \ 0]$ .
- 3) No super polynomial over  $F$  which annihilate  $T$  has smaller degree than  $P$ .

In the case of square super diagonal square matrices  $A$ , we define the minimal super polynomial for  $A$  as the unique monic super generator of super ideal of all super polynomials over  $F$  which super annihilate  $A$ .

If the operator  $T$  represented in some ordered super basis by the super square diagonal square matrix  $A$ , then  $T$  and  $A$  have the same minimal super polynomial, that is because

$$f(T) = [f_1(T_1) \ f_2(T_2) \ \cdots \ f_n(T_n)]$$

is represented by the matrix

$$f(A) = [f_1(A_1) \ f_2(A_2) \ \cdots \ f_n(A_n)]$$

so that  $f(T) = [0 \ 0 \ \cdots \ 0]$  if and only

$$\text{if } f(A) = [0 \ 0 \ \cdots \ 0]$$

(i.e :  $f_i(T_i) = 0$  if and only if  $f_i(A_i) = 0$ ).

Theorem (3.1) : Let  $T$  be a linear operator on an  $(n_1, n_2, \dots, n_n)$  dimensional vector space  $V = [V_1 \ V_2 \ \cdots \ V_n]$  or ( let  $A$  be an  $(n_1 \times n_1, n_2 \times n_2, \dots, n_n \times n_n)$  square super diagonal square matrix ). The characteristic super polynomial and minimal super polynomials for  $T$  (for  $A$ ) have the same super roots .

Proof : Let  $P = [P_1 \ P_2 \ \cdots \ P_n]$  be the minimal super polynomial for

$$T = [T_1 \ T_2 \ \cdots \ T_n]$$

( i.e :  $P_i$  is the minimal polynomial of  $T_i$ ,  $i = 1, 2, \dots, n$  ).

Let  $c = (c_1 \ c_2 \ \cdots \ c_n)$  be a scalar . we want to prove

$$P(c) = [P(c_1) \ P(c_2) \ \cdots \ P(c_n)] = [0 \ 0 \ \cdots \ 0]$$

if and only if  $c$  is the characteristic super value for  $T$ .

First we suppose

$$p(c) = [P_1(c_1) \ P_2(c_2) \ \cdots \ P_n(c_n)] = [0 \ 0 \ \cdots \ 0]$$

Then

$$P = [P_1 \ P_2 \ \cdots \ P_n] = (x - c)q = [(x - c_1)q_1 \ (x - c_2)q_2 \ \cdots \ (x - c_n)q_n]$$

where  $q = [q_1 \ q_2 \ \cdots \ q_n]$  is a super polynomial, since super degree of  $q$  is less than the super degree of  $P$ .

(i.e  $\deg(q_i) < \deg(P_i)$   $i = 1, 2, \dots, n$ ). the definition of the minimal super polynomial  $P$  tells us that

$$q(T) = [q_1(T_1) \ q_2(T_2) \ \cdots \ q_n(T_n)] \neq [0 \ 0 \ \cdots \ 0]$$

Now choose a super vector

$$\beta = [\beta_1 \ \beta_2 \ \cdots \ \beta_n] \text{ such that}$$

$$q(T)\beta = \begin{bmatrix} q_1(T_1)\beta_1 & q_2(T_2)\beta_2 & \cdots & q_n(T_n)\beta_n \end{bmatrix} = \begin{bmatrix} 0 & 0 & \cdots & 0 \end{bmatrix}$$

Now let

$$\alpha = \begin{bmatrix} \alpha_1 & \alpha_2 & \cdots & \alpha_n \end{bmatrix} = q(T)\beta = \begin{bmatrix} q_1(T_1)\beta_1 & q_2(T_2)\beta_2 & \cdots & q_n(T_n)\beta_n \end{bmatrix}$$

Then

$$\begin{bmatrix} 0 & 0 & \cdots & 0 \end{bmatrix} = P(T)\beta = \begin{bmatrix} P_1(T_1)\beta_1 & P_2(T_2)\beta_2 & \cdots & P_n(T_n)\beta_n \end{bmatrix}$$

$$= \begin{bmatrix} (T_1 - c_1 I_1)q_1(T_1)\beta_1 & (T_2 - c_2 I_2)q_2(T_2)\beta_2 & \cdots & (T_n - c_n I_n)q_n(T_n)\beta_n \end{bmatrix}$$

$$= \begin{bmatrix} (T_1 - c_1 I_1)\alpha_1 & (T_2 - c_2 I_2)\alpha_2 & \cdots & (T_n - c_n I_n)\alpha_n \end{bmatrix}$$

$$= (T - cI)\alpha.$$

And this  $c$  is a characteristic super value of  $T$ .

Conversely suppose that  $c$  is a characteristic super value of  $T$ . say

$$\begin{aligned} T(\alpha) &= c(\alpha) \\ &= \begin{bmatrix} T_1(\alpha_1) & T_2(\alpha_2) & \cdots & T_n(\alpha_n) \end{bmatrix} \\ &= \begin{bmatrix} c_1 \alpha_1 & c_2 \alpha_2 & \cdots & c_n \alpha_n \end{bmatrix} \end{aligned}$$

With

$$\alpha = \begin{bmatrix} \alpha_1 & \alpha_2 & \cdots & \alpha_n \end{bmatrix} \neq \begin{bmatrix} 0 & 0 & \cdots & 0 \end{bmatrix}$$

$$P(T)\alpha = \begin{bmatrix} P_1(T_1)\alpha_1 & P_2(T_2)\alpha_2 & \cdots & P_n(T_n)\alpha_n \end{bmatrix}$$

$$= \begin{bmatrix} P_1(c_1)\alpha_1 & P_2(c_2)\alpha_2 & \cdots & P_n(c_n)\alpha_n \end{bmatrix}$$

$$= P(c)\alpha$$

But  $P(T) = \begin{bmatrix} 0 & 0 & \cdots & 0 \end{bmatrix}$  then

$$P(T)\alpha = \begin{bmatrix} 0 & 0 & \cdots & 0 \end{bmatrix} = P(c)\alpha$$

So  $P(c) = 0$

Theorem(3.2): ( Cayley -Hamilton ) :

If  $A$  is a given  $n \times n$  matrix and  $P(\lambda)$  is the characteristic polynomial of  $A$  defined as :

$$P(\lambda) = \det (\lambda I_n - A)$$

Then  $P(A) = 0$

Proof : Let  $B = \text{adj} (tI_n - A)$

Then according to the right hand fundamental relation of adjugate one has

$$(tI_n - A) \cdot B = \det (tI_n - A)I_n = P(t)I_n$$

Since  $B$  is also a matrix with polynomials in  $t$  as entries , one can for each  $i$  collect the coefficients of  $t^i$  in each entry to form a matrix  $B_i$  of numbers , such that one has

$$B = \sum_{i=0}^{n-1} t^i B_i$$

(The way the entries of  $B$  are defined makes clear that no powers higher than  $t^{n-1}$  occur). While this looks like a polynomial with matrices as coefficients, we shall not consider such a notion; it is just a way to write a matrix with polynomial entries as linear combination of constant matrices, and the coefficient  $t^i$  has been written to the left of the matrix to stress this point of view. Now one can expand the matrix product in our equation by bilinearity :

$$P(t)I_n = (tI_n - A) \cdot B$$

$$= (tI_n - A) \cdot \sum_{i=0}^{n-1} t^i B_i$$

$$= \sum_{i=0}^{n-1} tI_n \cdot t^i B_i - \sum_{i=0}^{n-1} A \cdot t^i B_i$$

$$= \sum_{i=0}^{n-1} t^{i+1} B_i - \sum_{i=0}^{n-1} t^i A \cdot B_i$$

$$= t^n B_{n-1} + \sum_{i=0}^{n-1} t^i (B_{i-1} - A \cdot B_i) - A \cdot B_0$$

Writing

$$P(t)I_n = t^n I_n + t^{n-1} c_{n-1} I + \cdots + t c_1 I_n + c_0 I_n$$

One obtains an equality of two matrices with polynomial entries, written as linear combinations of constant matrices with powers of  $t$  as coefficients. Such an equality can hold only if in any matrix position the entry that is multiplied by a given power  $t^i$  is the same on both sides; it follows that the constant matrices with coefficient  $t^i$  in both expressions must be equal. Writing these equations for  $i$  from  $n$  down to 0 one finds :

$$B_{n-1} = I_n, \quad B_{i-1} - A \cdot B_i = c_i I_n \\ \text{for } 0 < i < n, \quad -AB_0 = c_0 I_n$$

We multiply the equation of the coefficients of  $t^i$  from the left by  $A_i$ , and sum up; the left-hand sides form a telescoping sum and cancel completely, which results in the equation :

$$0 = A^n + c_{n-1} A^{n-1} + \cdots + c_1 A + c_0 I_n = P(A)$$

This completes the proof.

**Example(3.8)** :Consider the super matrix :

$$A = \left[ \begin{array}{cc|cc} 1 & 2 & 0 & 0 \\ 3 & 4 & 0 & 0 \\ \hline 0 & 0 & 1 & 1 \\ 0 & 0 & 2 & 1 \end{array} \right] = \left[ \begin{array}{cc} A_1 & 0 \\ 0 & A_2 \end{array} \right]$$

Then the super characteristic super polynomial is :

$$P(\lambda) = [\lambda^2 - 5\lambda - 2 \mid \lambda^2 - 2\lambda - 1]$$

and

$$P(A) = [A_1^2 - 5A_1 - 2I \mid A_2^2 - 2A_2 - I] \\ = \left[ \begin{array}{cc|cc} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right] = [0 \mid 0]$$

### PART THREE

#### APPLICATION FOR BVP

We will consider the use of block elimination for the calculation of generalized turning and bifurcation points for two point B.V.P's. It will be shown that such algorithm will reduce the amount of work required in terms of LU-

factorizations to minimal. Since the discretization error of the approximated solution has an asymptotic expansion in terms of even powers of  $h$  (the step size). This will lead to the use of some type of extrapolation to produce more accurate results. Finite differences will be used to discretize the two point B.V.P's.

The following examples were used for numerical experimentation. We solved the one-dimensional nonlinear problem :

$$y'' + \lambda e^y = 0 \text{ on the interval } [0,1]$$

With the boundary conditions

$y(1) = y(0) = 0$  which has a simple turning point at the critical parameter ;  $\lambda = 3.513807$  with the initial guess  $\lambda = 3.4$ . The results for the fifth iteration with

$$h = \frac{1}{3}, \frac{1}{6}, \frac{1}{12}, \text{ and } \frac{1}{24} \text{ using block}$$

elimination are given in Table 1.

**Table 1**

$h$	$g$	$\lambda$
$\frac{1}{3}$	0.143051E-05	3.31092
$\frac{1}{6}$	-0.722452E-05	3.46261
$\frac{1}{12}$	0.560958E-05	3.50110
$\frac{1}{24}$	0.471423E-05	3.51062

Using the results in Table 1, we applied the Richardson extrapolation and obtained  $\lambda = 3.51378$ . The results of the application of the extrapolation are given in Table 2.

**Table 2.**

$h$	Number of Extrapolation			
	0	1	2	3
$\frac{1}{3}$	3.31092			
		3.51318		
$\frac{1}{6}$	3.46261		3.51398	
		3.51393		3.51378
$\frac{1}{12}$	3.50110		3.51378	
		3.51379		
$\frac{1}{24}$	3.51062			

It is clear from Table 1 that  $g \rightarrow 0$  for the various values of  $h$  as expected . The CPU time is equal to 0.040 sec .

Repeating the same calculations for

$$h = \frac{1}{3}, \frac{1}{6}, \frac{1}{12}, \text{ and } \frac{1}{24} \text{ but without using the}$$

block-elimination this time . The results were almost the same are given in Table 3 .

**Table 3**

$h$	$g$	$\lambda$
$\frac{1}{3}$	0.1172142E-03	3.31091
$\frac{1}{6}$	.395983E-05	3.46261
$\frac{1}{12}$	.171911E-05	3.50110
$\frac{1}{24}$	-.404437E-05	3.51062

Again applying Richardson extrapolation on approximate values of  $\lambda_h$  the results are given in Table 4 .

**Table 4.**

$h$	Number of Extrapolation			
	0	1	2	3
$\frac{1}{3}$	3.31092			
		3.51318		
$\frac{1}{6}$	3.46261		3.51398	
		3.51393		3.51378
$\frac{1}{12}$	3.50110		3.51378	
		3.51379		
$\frac{1}{24}$	3.51062			

The CPU time is equal to 0.240 sec . It clear that the CPU time used with block elimination is  $\frac{1}{6}$  of the time used without block elimination .

We also solved the system with  $h = \frac{1}{50}$  using

block elimination . With the same initial guess

$\lambda = 3.4$  , we obtain the solution

$\lambda = 3.51295$  and  $g = 0.98533E - 04$  .

The CPU time was 0.190 sec . This shows

that with such a large system , we were not able to obtain the same accuracy as we obtained in Table 3, also the time required is 4 times that needed with Richardson extrapolation .

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