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# **MATH 537, Fall 2020**

## **Ordinary Differential Equations**

Lecture #16: Part II

Chapter 5  
Higher-Dimensional Linear Algebra

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## Section 5.2

# A Summary For Section 5.2: Eigenvalues

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- Eigenvalue Problems

$$AV = \lambda V \Rightarrow (A - \lambda I)V = 0 \Rightarrow |A - \lambda I| = 0$$

- Linearly Independent eigenvectors associated with real and distinct eigenvalues

$$\det|B| \neq 0$$

$B$  consists of linearly independent vectors

- Diagonalization

$$T = (V_1, V_2, \dots, V_n), \quad V_j \text{ are eigenvectors}$$

$$T^{-1}AT = D$$

## 5.2 Eigenvalue Problem and LI Eigenvectors

### Definition

A vector  $V$  is an *eigenvector* of an  $n \times n$  matrix  $A$  if  $V$  is a nonzero solution to the system of linear equations  $(A - \lambda I)V = 0$ . The quantity  $\lambda$  is called an *eigenvalue* of  $A$ , and  $V$  is an eigenvector associated to  $\lambda$ .

$$AV = \lambda V \quad \Rightarrow \quad (A - \lambda I)V = 0 \quad \Rightarrow \quad |A - \lambda I| = 0$$

**Proposition.** Suppose  $\lambda_1, \dots, \lambda_\ell$  are real and distinct eigenvalues for  $A$  with associated eigenvectors  $V_1, \dots, V_\ell$ . Then the  $V_j$  are linearly independent. ■

$V_j$  are Linearly Independent

## Sect 5.2 Diagonalization

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**Corollary.** Suppose  $A$  is an  $n \times n$  matrix with real, distinct eigenvalues. Then there is a matrix  $T$  such that

$$T^{-1}AT = \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix} = D$$

where all of the entries off the diagonal are 0.

For example,

$T = (V_1, V_2, \dots, V_n)$ ,  $V_j$  are eigenvectors

## Sect 5.2 Eigenvalues and Eigenvectors

**Example.** Let

$$A = \begin{pmatrix} 1 & 2 & -1 \\ 0 & 3 & -2 \\ 0 & 2 & -2 \end{pmatrix}.$$

$$AV = \lambda V$$

$$A - \lambda I = \begin{pmatrix} 1 - \lambda & 2 & -1 \\ 0 & 3 - \lambda & -2 \\ 0 & 2 & -2 - \lambda \end{pmatrix}$$

$$|A - \lambda I| = 0$$

$$(1 - \lambda)(\lambda^2 - \lambda - 2) = 0$$

$$\lambda = 2, 1, -1$$

$$\lambda_1 = 2$$

$$\lambda_2 = 1$$

$$\lambda_3 = -1$$

$$V_1 = \begin{pmatrix} 3 \\ 2 \\ 1 \end{pmatrix}$$

$$V_2 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$$

$$V_3 = \begin{pmatrix} 0 \\ 1 \\ 2 \end{pmatrix}$$

## Construct $T$ and $D$

Let  $D = \begin{pmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{pmatrix}$   $\lambda_j$  are eigenvalues of the matrix  $A$

Let  $T = (V_1, V_2, V_3)$ ,  $V_j$  are the eigenvectors corresponding to  $\lambda_j$

$$\lambda_1 = 2$$

$$\lambda_2 = 1$$

$$\lambda_3 = -1$$

$$V_1 = \begin{pmatrix} 3 \\ 2 \\ 1 \end{pmatrix}$$

$$V_2 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$$

$$V_3 = \begin{pmatrix} 0 \\ 1 \\ 2 \end{pmatrix}$$

$$D = \begin{pmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{pmatrix}$$

$$T = (V_1, V_2, V_3) = \begin{pmatrix} 3 & 1 & 0 \\ 2 & 0 & 1 \\ 1 & 0 & 2 \end{pmatrix}$$

## Diagnoalization using $T^{-1}AT$

$$\text{Compute } AT = \begin{pmatrix} 1 & 2 & -1 \\ 0 & 3 & -2 \\ 0 & 2 & -2 \end{pmatrix} \begin{pmatrix} 3 & 1 & 0 \\ 2 & 0 & 1 \\ 1 & 0 & 2 \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix}$$

- Pick one and compute
  - During the difficult time, the following is added to make our life easier:
    - you will receive additional 10 points for your next homework if your selection matches the one I preselected (to appear later) and your answer is correct.
  - Send your results via "chat" (e.g., a21=???)
  - You have 2 minutes
- 
- You are a winner if you select  $a_{33}$  and compute it correctly.

$$AT = \begin{pmatrix} 1 & 2 & -1 \\ 0 & 3 & -2 \\ 0 & 2 & -2 \end{pmatrix} \begin{pmatrix} 3 & 1 & 0 \\ 2 & 0 & 1 \\ 1 & 0 & 2 \end{pmatrix} = \begin{pmatrix} 6 & 1 & 0 \\ 4 & 0 & -1 \\ 2 & 0 & -2 \end{pmatrix}$$



## Diagnoalization using $T^{-1}AT$

$$AT = \begin{pmatrix} 1 & 2 & -1 \\ 0 & 3 & -2 \\ 0 & 2 & -2 \end{pmatrix} \begin{pmatrix} 3 & 1 & 0 \\ 2 & 0 & 1 \\ 1 & 0 & 2 \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix} = \begin{pmatrix} 6 & 1 & 0 \\ 4 & 0 & -1 \\ 2 & 0 & -2 \end{pmatrix}$$

What we have done is called “parallel computing”:

- Compute each of  $a_{ij}$  in parallel
- Load imbalance may appear, e.g.,  $a_{12} = 1 + 2 * 0 + (-1) * 0$
- The last element determines the timing for the entire task.

In the supercomputing world, we may additionally perform the following:

- **Decompose** data (domain or tasks) into smaller pieces of data;
- Assign sub-tasks (e.g., **broadcast** data) to different CPUs;
- **Gather** all of sub-tasks and put them together

P1

```
for i in range (0, N/2):  
    a[i]=c*b[i]
```

P2

```
for i in range (N/2, N):  
    a[i]=c*b[i]
```

```
N=1000  
N2=N/2  
a=np.zeros(N)  
b=np.linspace(0, N-1, N)  
c=5
```

```
for i in range (0, N):  
    a[i]=c*b[i]
```

```
x1 = np.linspace(0, 10, N, endpoint=True)  
x2 = np.linspace(0, 10, N, endpoint=False)
```

```

xsum = 0.0
do i=1,15
  xsum = xsum + x(i)
enddo

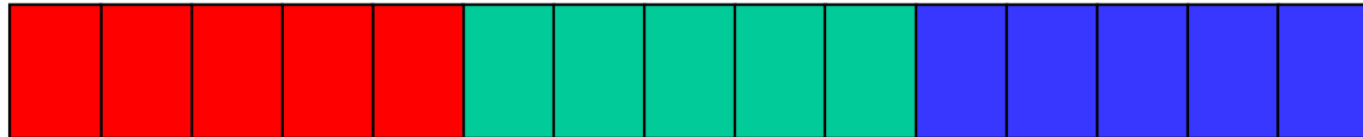
```

```

real x(15)

```

“Global”    1   2   3   4   5   6   7   8   9   10   11   12   13   14   15



“Local”    1   2   3   4   5   1   2   3   4   5   1   2   3   4   5

P1

P2

P3

**real x(15)**

1	2	3	4	5	6	7	8	9	10	11	12	13	14	15

## Parallel Code

**real x(5)**

--	--	--	--	--

**1 2 3 4 5**

**P1**

**real x(5)**

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**1 2 3 4 5**

**P2**

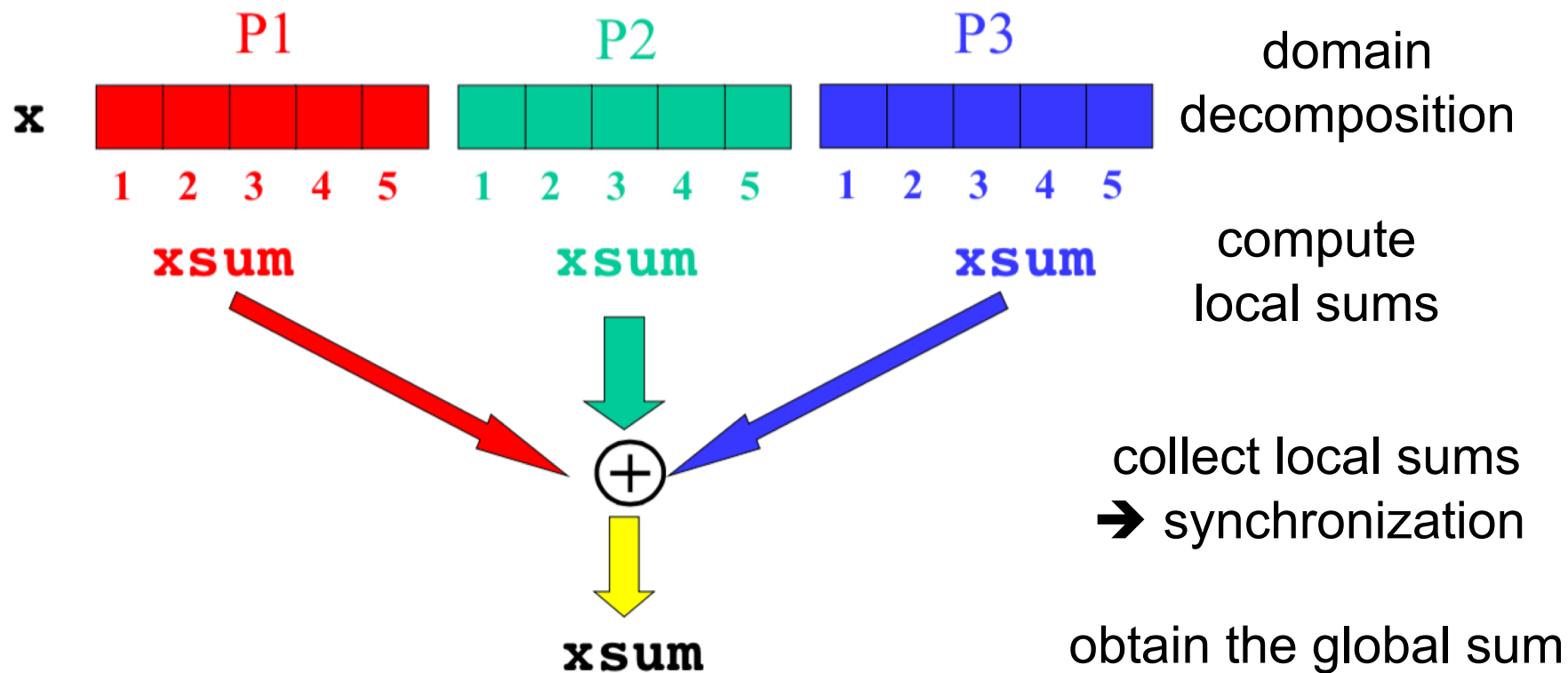
**real x(5)**

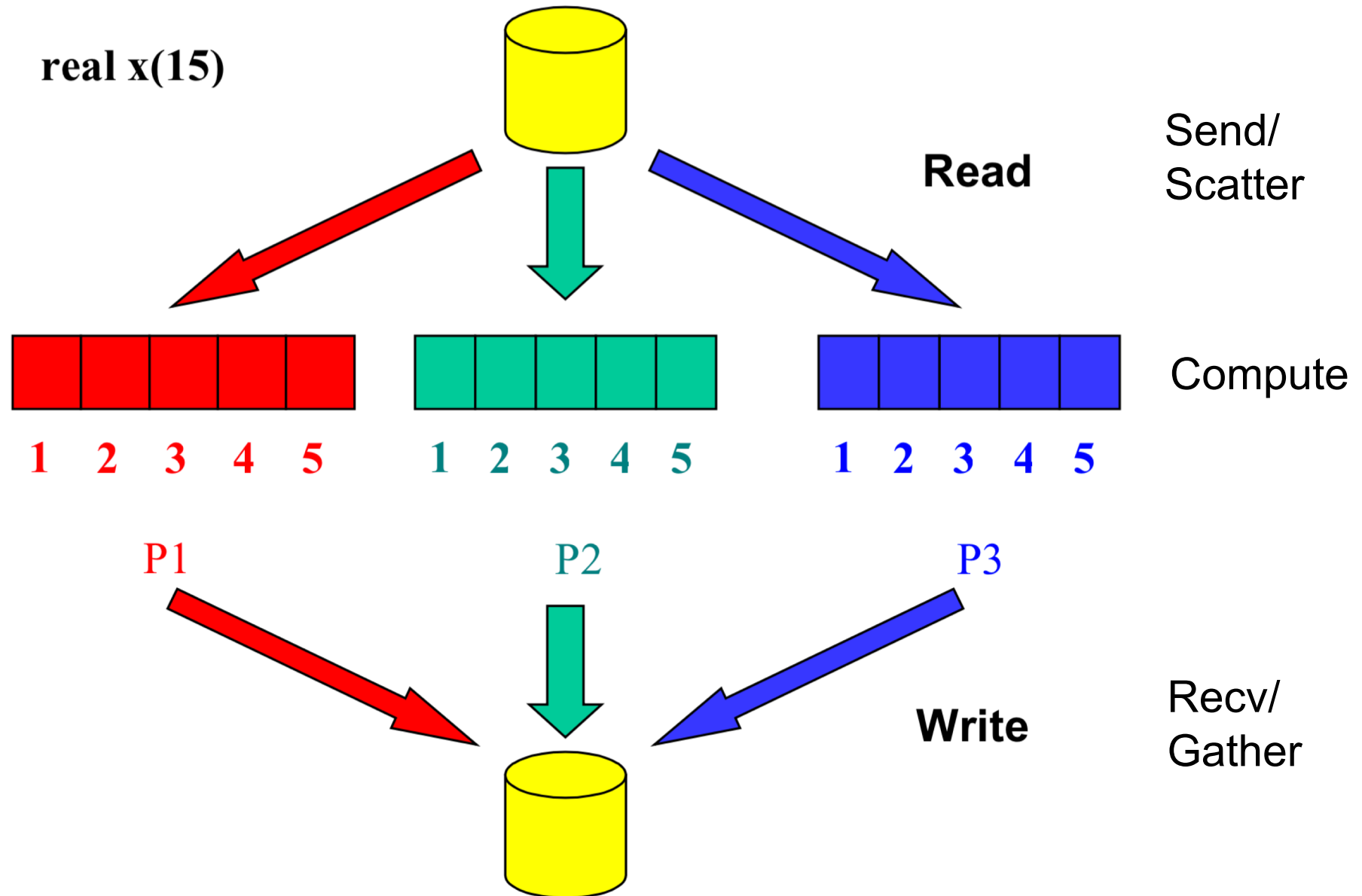
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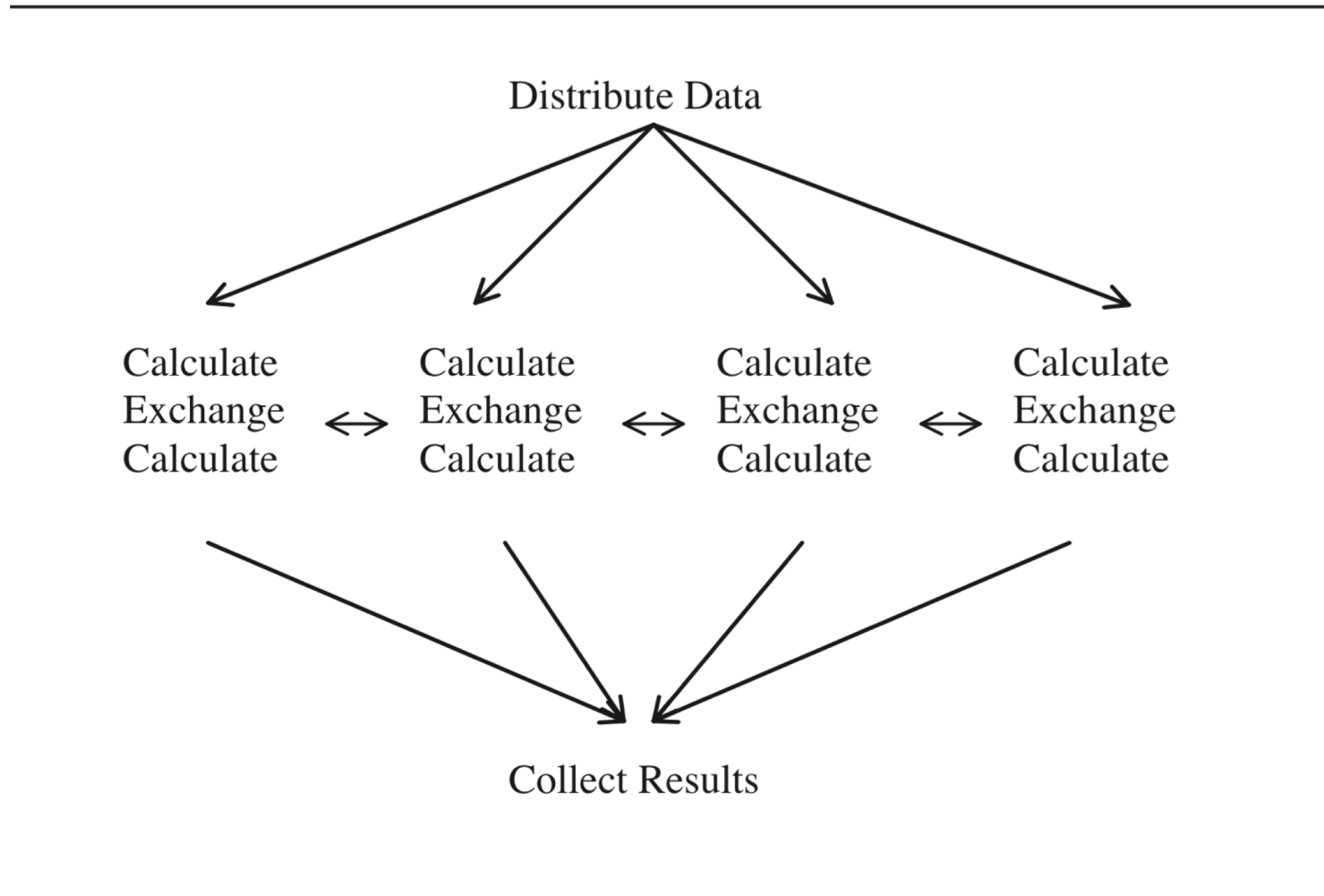
**1 2 3 4 5**

**P3**

```
xsum = 0.0
do i=1,15
  xsum = xsum + x(i)
enddo
```







**Figure 1.5** Basic structure of a SPMD program.

# A Minimal Set for MPI

A minimal set of routines that most parallel codes run with are:

- **MPI\_INIT:**  
Initialization. MPI spawns an identical copy of my\_proc when `mpirun -np N my_proc` is issued.
- **MPI\_COMM\_SIZE:**  
returns the **size** or number (i.e., **N**) of processes in the application.
- **MPI\_COMM\_RANK:**  
returns the rank ("**id**", **0~N-1**) of the calling process.
- **MPI\_SEND**
- **MPI\_RECV**
- **MPI\_WAIT**
- **MPI\_FINALIZE:**  
terminates all MPI processing



# A Minimal Set for MPI in Python

A minimal set of routines that most parallel codes run with are:

	Fortran	MPI4PY	
	MPI_INIT	-----	comm = MPI.COMM_WORLD
	MPI_COMM_SIZE	comm.Get_size()	
	MPI_COMM_RANK	comm.Get_rank()	
	MPI_SEND	comm.send(...)	comm.Send(...)
	MPI_RECV	comm.recv(...)	comm.Recv(...)
	MPI_WAIT	obj.wait()	
	MPI_FINALIZE	-----	

## Diagonalization using $T^{-1}AT$

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$$AT = \begin{pmatrix} 1 & 2 & -1 \\ 0 & 3 & -2 \\ 0 & 2 & -2 \end{pmatrix} \begin{pmatrix} 3 & 1 & 0 \\ 2 & 0 & 1 \\ 1 & 0 & 2 \end{pmatrix} = \begin{pmatrix} 6 & 1 & 0 \\ 4 & 0 & -1 \\ 2 & 0 & -2 \end{pmatrix}$$

$$TD = \begin{pmatrix} 3 & 1 & 0 \\ 2 & 0 & 1 \\ 1 & 0 & 2 \end{pmatrix} \begin{pmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{pmatrix} = \begin{pmatrix} 6 & 1 & 0 \\ 4 & 0 & -1 \\ 2 & 0 & -2 \end{pmatrix}$$

$$TD = AT$$

$$T^{-1}TD = T^{-1}AT$$

$$ID = T^{-1}AT$$

$$D = T^{-1}AT$$

## Section 5.3 Complex Eigenvalues

if  $AV = (\alpha + i\beta)V$ , we have  $A\bar{V} = (\alpha - i\beta)\bar{V}$

- Let  $V$  be the eigenvector associated with the complex eigenvalue  $\alpha + i\beta$
  - Show that  $\bar{V}$  is an eigenvector associated with the complex eigenvalue  $\alpha - i\beta$
  - $V$  and  $\bar{V}$  are linearly independent, i.e.,  $cV + d\bar{V} = 0 \iff c = d = 0$ 
    - $c$  &  $d$  are complex numbers.
- Note that  $V$  and  $\bar{V}$  yield the same independent real functions, because  $\text{Re}(V) = \text{Re}(\bar{V})$  and  $\text{Im}(V) = -\text{Im}(\bar{V})$ .
- From the first bullet, we have  $AV = (\alpha + i\beta)V$ .
- To prove the statement in the 2<sup>nd</sup> bullet, we consider  $A\bar{V}$ :

$$A\bar{V} = \overline{AV} = \overline{(\alpha + i\beta)V} = (\alpha - i\beta)\bar{V}$$

# $Re(V)$ and $Im(V)$ Are Linearly Independent

Supp

$$V = u + i w$$

Show that  $u$  and  $w$  are LI

$$\text{Assume } w = cu, c \in R$$

$$\lambda = \alpha + i\beta$$

$$AV = A(u + iw) = A(u + icu) \quad (1)$$

$$\lambda V = (\alpha + i\beta)(u + icu) = (\alpha u - c\beta u) + i(\beta u + c\alpha u) \quad (2)$$

Eq. (1) = (2):

$$\text{real part: } Au = \alpha u - c\beta u \quad (3)$$

$$\text{imaginary part: } Acu = \beta u + c\alpha u \quad (4)$$

$$(3)^*c = (4),$$

$$-c^2\beta u = \beta u$$

$$c^2 = -1, \quad c = \pm i$$

contradiction

# Construct a Linear Map, $T$ , using Real and Imaginary Parts

Assume  $A$  to be a  $(2n \times 2n)$  matrix that has the following eigenvalues and eigenvectors:

- $\alpha_j \pm \beta_j$  and  $V_j, \bar{V}_j, j = 1, 2$ ;
- $AV_j = (\alpha_j + i\beta)V_j$  and  $A\bar{V}_j = (\alpha_j - i\beta)\bar{V}_j$

Define the following

$$W_{2j-1} = \frac{1}{2}(V_j + \bar{V}_j) = \text{Re}(V_j)$$

$$W_{2j} = \frac{-i}{2}(V_j - \bar{V}_j) = \text{Im}(V_j)$$

Show that (TBD in the next slide)

$$AW_{2j-1} = \alpha_j W_{2j-1} - \beta_j W_{2j}$$

$$AW_{2j} = \beta_j W_{2j-1} + \alpha_j W_{2j}$$

Construct  $T$  as follows:

$$T = [W_1, W_2, \dots, W_{2j-1}, W_{2j} \dots W_{2n-1}, W_{2n}]$$

Obtain  $Y' = BY, X = TY$ , and  $B$  is defined as follows:

$$B = T^{-1}AT$$

## Find $AW_{2j-1}$ and $AW_{2j}$

Show that

$$AW_{2j-1} = \alpha_j W_{2j-1} - \beta_j W_{2j} \quad AW_{2j} = \beta_j W_{2j-1} + \alpha_j W_{2j}$$

$$AW_{2j-1} = \frac{1}{2} (AV_j + A\bar{V}_j) = \frac{1}{2} ((\alpha_j + i\beta)V_j + (\alpha_j - i\beta)\bar{V}_j)$$

$$= \frac{1}{2} (\alpha_j(V_j + \bar{V}_j) + i\beta_j(V_j - \bar{V}_j)) = \alpha_j W_{2j-1} - \beta_j W_{2j}$$

$$AW_{2j} = \frac{-i}{2} (AV_j - A\bar{V}_j) = \frac{-i}{2} ((\alpha_j + i\beta)V_j - (\alpha_j - i\beta)\bar{V}_j)$$

$$= \frac{-i}{2} (\alpha_j(V_j - \bar{V}_j) + i\beta_j(V_j + \bar{V}_j)) = \beta_j W_{2j-1} + \alpha_j W_{2j}$$

$W_1, W_2, \dots, W_{2j-1}, W_{2j}, \dots, W_{2n-1}, W_{2n}$  are LI

$$W_{2j-1} = \frac{1}{2}(V_j + \bar{V}_j) = \text{Re}(V_j)$$

$$W_{2j} = \frac{-i}{2}(V_j - \bar{V}_j) = \text{Im}(V_j)$$

**Proposition.** *The vectors  $W_1, \dots, W_{2n}$  are linearly independent.*

Form a linear combination:

$$\sum_{j=1}^n (c_j W_{2j-1} + d_j W_{2j}) = 0$$

They are LD for **some non-zero**  $c_j, d_j \in \mathbb{R}$

Plug the Eqs. on the top into the above Eq.:

$$\frac{1}{2} \sum_{j=1}^n (c_j - id_j)V_j + (c_j + id_j)\bar{V}_j = 0$$

Since the  $V_j$  and the  $\bar{V}_j$  are LI, we have

$$(c_j - id_j) = 0 = (c_j + id_j) \Rightarrow c_j = d_j = 0 \text{ contradiction}$$

# Construct a Linear Map, $T$

$$AW_{2j-1} = \alpha_j W_{2j-1} - \beta_j W_{2j} \quad AW_{2j} = \beta_j W_{2j-1} + \alpha_j W_{2j}$$

Construct  $T = [W_1, W_2, \dots, W_{2j-1}, W_{2j}, \dots, W_{2n-1}, W_{2n}]$

Obtain  $TE_j = W_j, \quad j = 1 \sim 2n, \quad T^{-1}W_j = E_j, \quad j = 1 \sim 2n,$

for  $2j-1$ ,  $(T^{-1}AT)E_{2j-1} = T^{-1}ATE_{2j-1} = T^{-1}AW_{2j-1}$

$$= T^{-1}(\alpha_j W_{2j-1} - \beta_j W_{2j})$$

$$= (\alpha_j T^{-1}W_{2j-1} - \beta_j T^{-1}W_{2j})$$

$$= (\alpha_j E_{2j-1} - \beta_j E_{2j})$$

for  $2j$ ,  $(T^{-1}AT)E_{2j} = (\beta_j E_{2j-1} + \alpha_j E_{2j})$



## Construct a Linear Map, $T$

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$$(T^{-1}AT)E_{2j-1} = (\alpha_j E_{2j-1} - \beta_j E_{2j})$$

$$(T^{-1}AT)E_{2j} = (\beta_j E_{2j-1} + \alpha_j E_{2j})$$

$j=1$ , we have

$$(T^{-1}AT)E_1 = (\alpha_1 E_1 - \beta_1 E_2) = \alpha_1 \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} - \beta_1 \begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix} = \begin{pmatrix} \alpha_1 \\ -\beta_1 \\ \vdots \\ 0 \end{pmatrix}$$

$$(T^{-1}AT)E_2 = (\beta_1 E_1 + \alpha_1 E_2) = \beta_1 \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} + \alpha_1 \begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix} = \begin{pmatrix} \beta_1 \\ \alpha_1 \\ \vdots \\ 0 \end{pmatrix}$$

## Construct a Linear Map, $T$

$$(T^{-1}AT)E_1 = (\alpha_1 E_1 - \beta_1 E_2) = \alpha_1 \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} - \beta_1 \begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix} = \begin{pmatrix} \alpha_1 \\ -\beta_1 \\ \vdots \\ 0 \end{pmatrix}$$

$$(T^{-1}AT)E_2 = (\beta_1 E_1 + \alpha_1 E_2) = \beta_1 \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} + \alpha_1 \begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix} = \begin{pmatrix} \beta_1 \\ \alpha_1 \\ \vdots \\ 0 \end{pmatrix}$$

$$(T^{-1}AT)[E_1, E_2, \dots, E_{2n-1}, E_{2n}] = \begin{pmatrix} \alpha_1 & \beta_1 \\ -\beta_1 & \alpha_1 \\ & & & \end{pmatrix}$$

# Construct a Linear Map, $T$

$$(T^{-1}AT)[E_1, E_2, \dots, E_{2n-1}, E_{2n}] = \begin{pmatrix} \alpha_1 & \beta_1 \\ -\beta_1 & \alpha_1 \end{pmatrix}$$

*Identity matrix  $I$*

$$T^{-1}AT = \begin{pmatrix} \alpha_1 & \beta_1 \\ -\beta_1 & \alpha_1 & & \\ & & \alpha_j & \beta_j \\ & & -\beta_j & \alpha_j \\ & & & & \ddots & \ddots \\ & & & & & \alpha_n & \beta_n \\ & & & & & -\beta_n & \alpha_n \end{pmatrix} \begin{matrix} \leftarrow 2j-1 \\ \leftarrow 2j \end{matrix}$$

$$D_1 = \begin{pmatrix} \alpha_1 & \beta_1 \\ -\beta_1 & \alpha_1 \end{pmatrix}$$

$$D_j = \begin{pmatrix} \alpha_j & \beta_j \\ -\beta_j & \alpha_j \end{pmatrix}$$

# A Summary with Complex Eigenvalues

Assume  $A$  to be a  $(2n \times 2n)$  matrix that has the following eigenvalues and eigenvectors:

- $\alpha_j \pm \beta_j$  and  $V_j, \bar{V}_j, j = 1, 2$ ;

Define the following

$$W_{2j-1} = \frac{1}{2}(V_j + \bar{V}_j) = \text{Re}(V_j) \quad W_{2j} = \frac{-i}{2}(V_j - \bar{V}_j) = \text{Im}(V_j)$$

Construct  $T$  as follows:

$$T = [W_1, W_2, \dots, W_{2j-1}, W_{2j} \dots W_{2n-1}, W_{2n}]$$

Obtain  $Y' = BY, X = TY$ , and  $B$  is defined as follows:

$$B = T^{-1}AT = \begin{pmatrix} D_1 & & \\ & \ddots & \\ & & D_n \end{pmatrix} \quad D_n = \begin{pmatrix} \alpha_n & \beta_n \\ -\beta_n & \alpha_n \end{pmatrix}$$