

MATLAB Parallel Computing Toolbox

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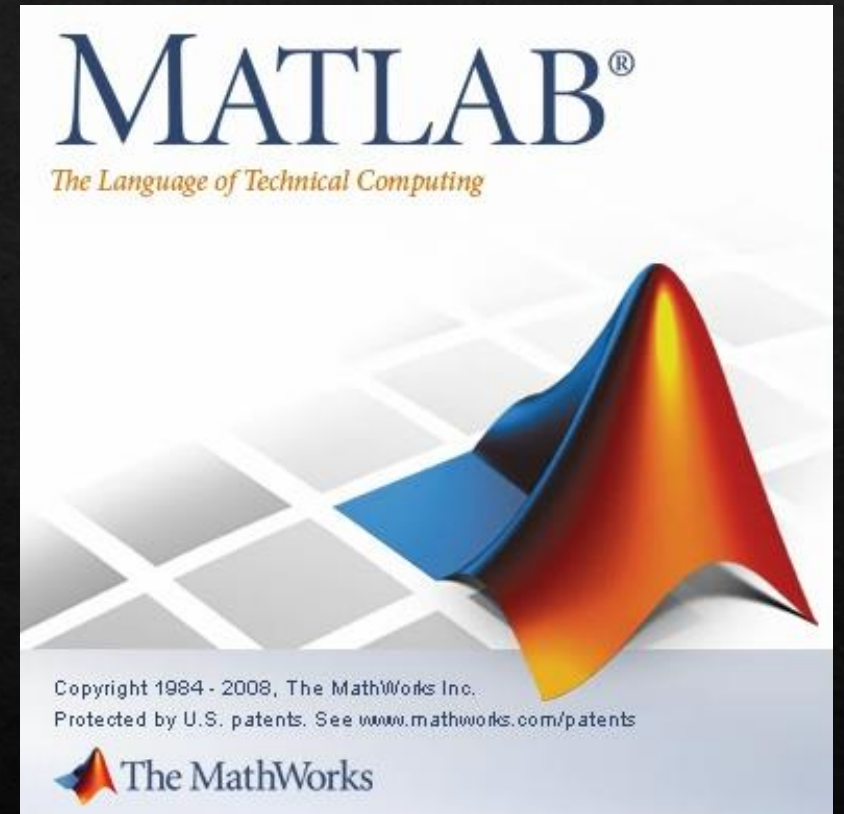
Boston University

Why using Matlab Parallel Computing Toolbox (PCT)?

- ◆ To accelerate a MATLAB program, it is necessary to parallelize it.
- ◆ Take advantages of high-performance computing (HPC) resources, such as multi-core CPUs, GPUs, and computer clusters.
- ◆ Boston University (BU) Shared Computing Cluster (SCC) is an HPC cluster with over 11,000 CPU processors and over 250 GPUs.
- ◆ MATLAB site license is available to all BU users. (Unlimited on SCC).
- ◆ The PCT can be used not only on HPC clusters or but also on regular laptops/desktops.
- ◆ The skills you learn today should enable you to solve bigger problems faster using MATLAB.

Outline

- ◇ Start up MATLAB on BU SCC
- ◇ Parallelize Matlab codes
 - ✓ Implicit parallelism
 - ✓ Explicit parallelism
 - ✓ Using GPU



Access to BU SCC resources

◇ Log in:

```
$ ssh -X username@scc2.bu.edu
```

◇ **Interactive session** (for working interactively on compute nodes):

```
$ qssh # Start an interactive session
```

```
$ qssh -pe omp 4 # Request 4 CPU cores
```

```
$ qssh -l gpus=1 # Request one GPU and one CPU core
```

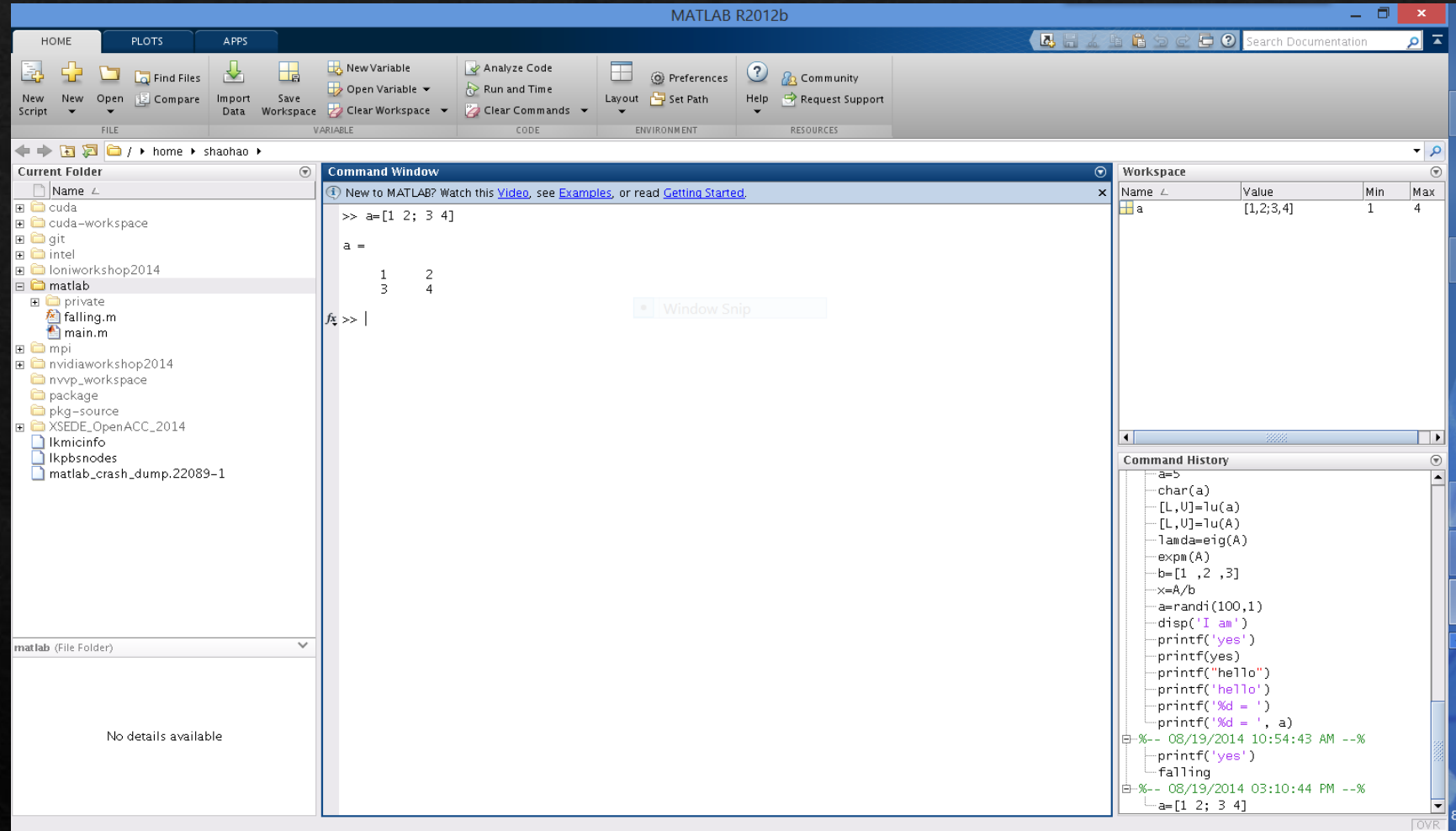
◇ **Use module to load Matlab:**

```
$ moduleavail | grep matlab # See all available versions
```

```
$ module load matlab # Set up environment variables
```

Graphic platform

□ Open Matlab
\$ matlab &

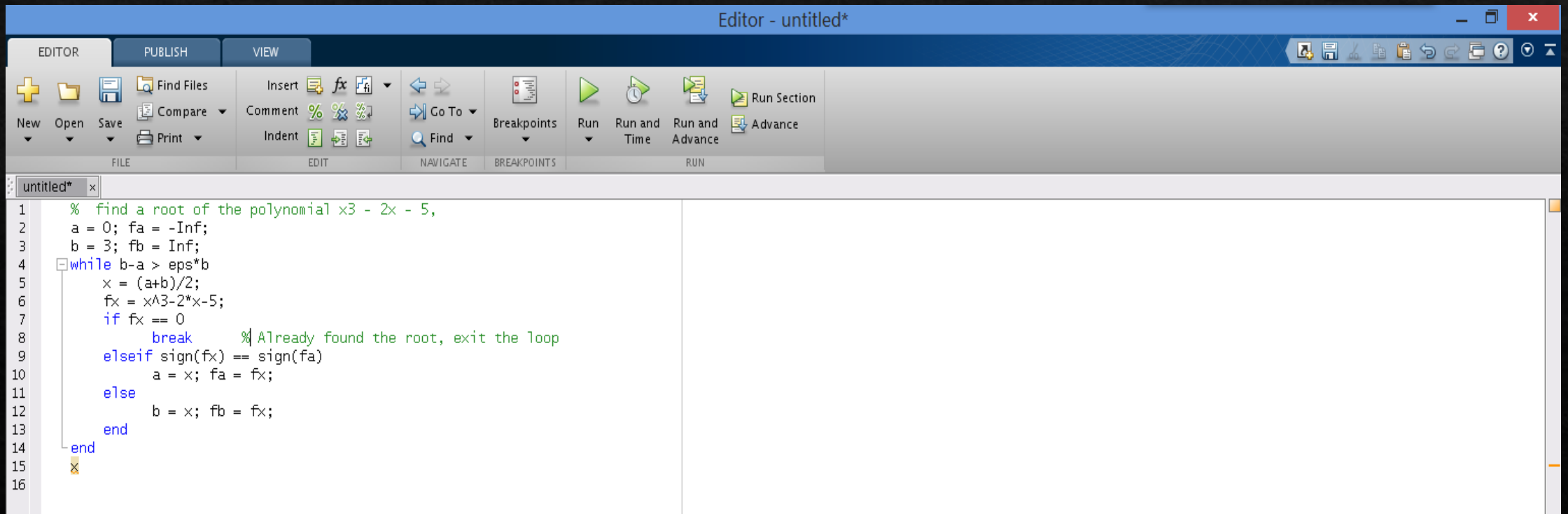


□ Use VNC to speed up graphical interface.

Refer to: <https://www.bu.edu/tech/support/research/system-usage/getting-started/remote-desktop-vnc/>

M-file

- ◇ An m-file is a simple text file where you can place MATLAB commands.
- ◇ Save your works
- ◇ Convenient for debugging
- ◇ Run directly. Pre-compiling is unnecessary.



The screenshot shows the MATLAB Editor interface with a script titled 'untitled*'. The script is designed to find a root of the polynomial $x^3 - 2x - 5$ using the bisection method. The code is as follows:

```
1 % find a root of the polynomial x3 - 2x - 5,  
2 a = 0; fa = -Inf;  
3 b = 3; fb = Inf;  
4 while b-a > eps*b  
5     x = (a+b)/2;  
6     fx = x3-2*x-5;  
7     if fx == 0  
8         break % Already found the root, exit the loop  
9     elseif sign(fx) == sign(fa)  
10        a = x; fa = fx;  
11    else  
12        b = x; fb = fx;  
13    end  
14 end  
15  
16
```

Text platform

`$ matlab -nodisplay` % Work in text interface. Does not display any graph.
`$ matlab -nodesktop` % Program in text interface. Pop out graphs when necessary.

❖ Many Linux commands (prefix an exclamation mark) are available within Matlab platform, such as:

`cd, ls, pwd, !cp, !rm, !mv, !cat, !vim, !diff, and !grep`

❖ Edit M-file and run the program:

`>> !vim mfilename.m` % edit in text window
`>> edit mfilename.m` % create a new or open an existing m-file in graphical window
`>> open mfilename.m` % open an existing m-file in graphical window
`>> run mfilename.m` % run the program
`>> mfilename` % run the program

Parallelize Matlab codes

- ◇ Parallel computing:
run multiple tasks by different workers simultaneously.
- ◇ Matlab parallel computing toolbox (PCT)
 - ✓ Implicit parallelism: automatic multi-threaded vector operations
 - ✓ Explicit parallelism: `parpool`, `parfor`, `spmd`
 - ✓ Using GPU: `gpuArray`, `arrayfun`

Parallel Computing

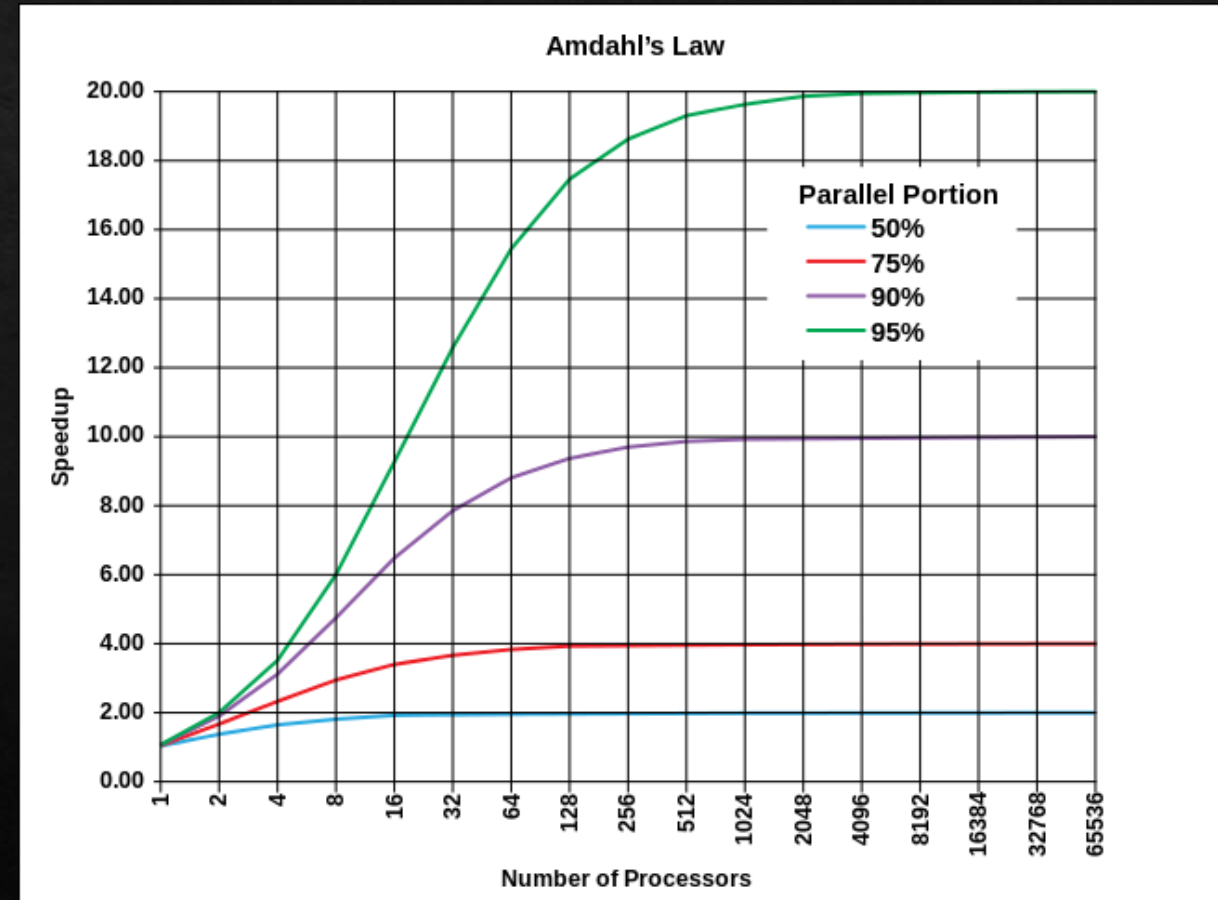
- Parallel computing is a type of computation in which many calculations are carried out **simultaneously**.

- **Speedup** of a parallel program,

$$S(p) = \frac{T(1)}{T(p)} = \frac{1}{\alpha + \frac{1}{p}(1 - \alpha)}$$

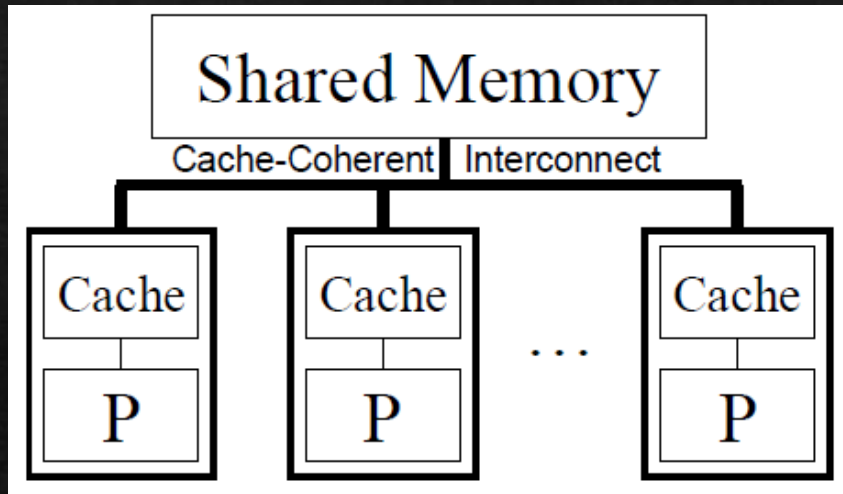
p : number of processors/cores,

α : fraction of the program that is serial.

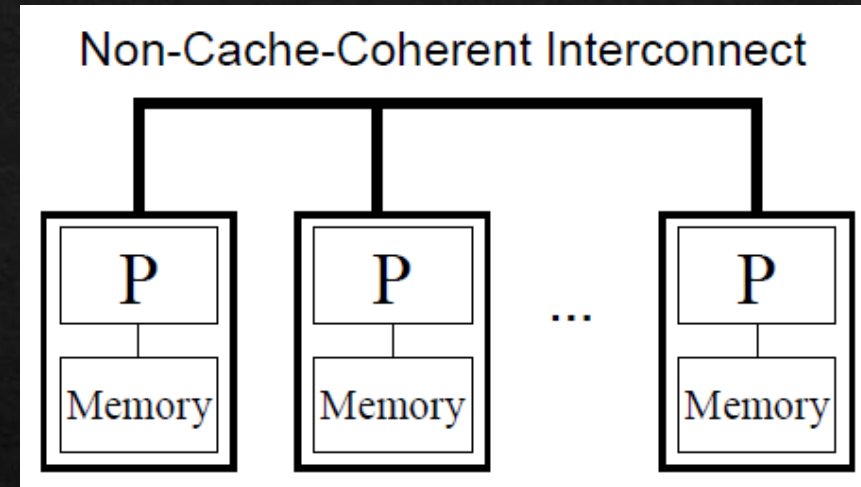


- Figure from: https://en.wikipedia.org/wiki/Parallel_computing

Two types of Parallel Computers



- Shared memory
- Multiple CPU cores within one node



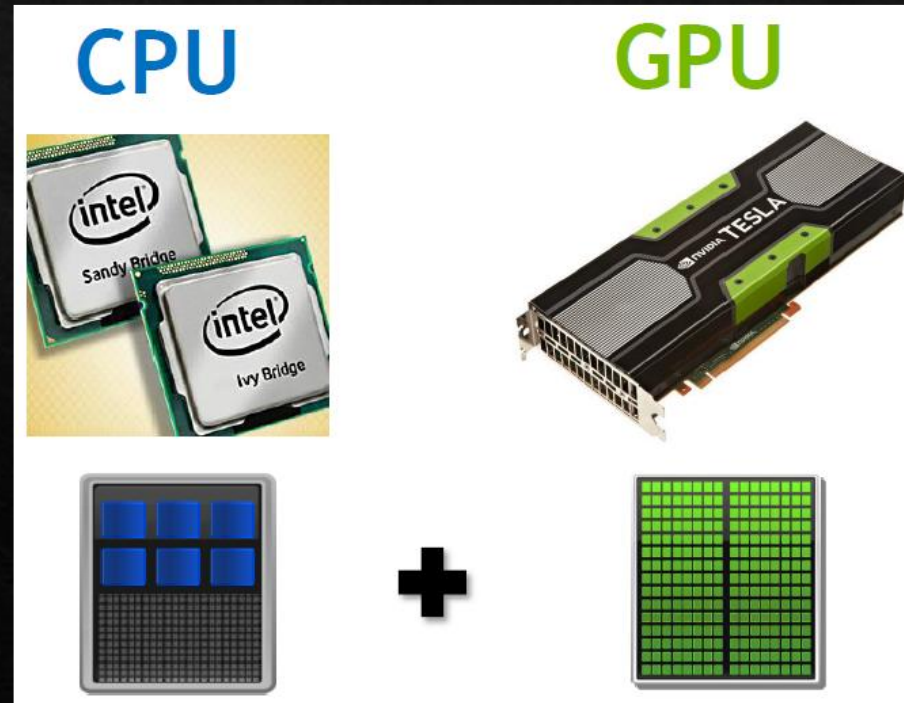
- Distributed memory
- Multiple nodes within one cluster

- ◇ Implicit (multithreaded) parallelism in Matlab is only for multiple cores on one node.
- ◇ Explicit parallelism in Matlab can be implemented on either single node or multiple nodes.

Only the single-node mode is supported on BU SCC currently.

Graphic Processing Unit (GPU)

- ◇ GPU is a device attached to a CPU-based system.
- ◇ Computer program can be parallelized and accelerated on GPU.
- ◇ CPU and GPU has separated memory. Data transfer between CPU and GPU is required.
- ◇ Many Matlab functions are enabled on GPU.



Implicit parallelism: multithreaded operations

◇ Many built-in operators or functions are implicitly multi-threaded, such as,

Basic: +, -, .*, ./, .^, *, ^, MAX, MIN, SUM, SORT, ABS

Elementary math: ATAN2, COS, CSC, SEC, SIN, TAN, EXP, POW2, SQRT, ABS, LOG, LOG10

Linear algebra: INV, LINSOLVE, \, LU, QR, EIG, SVD

Data analysis: FFT, CONV2

✓ Multithreading may be triggered for vector implementation but not for loop implementation.

◇ By default on BU SCC, these operations automatically use the requested number of CPU cores in a batch job.

◇ Use function `maxNumCompThreads(n)` to limit the number of cores to be used.

Multithreaded Matrix Multiplication

```
n=7000;  
  
A=randn(n); B=randn(n); % initialize data  
C=zeros(n); D=zeros(n);  
  
tic % start measuring time  
C = A * B; % multithreaded by default  
toc % end measuring time  
  
maxNumCompThreads(1); % enforce using 1 thread  
tic  
D = A * B; % single thread  
toc
```

parpool

- ◇ **parpool** enables the full functionality of the parallel language features (e.g. **parfor** and **spmd**) by creating a special job on a pool of **workers**, and connecting the MATLAB **client** to the parallel pool.
- ✓ **Client/master**: runs serial work. Interactive with users (e.g. for input, output, serial parts).
- ✓ **Workers/labs**: run parallel work. Typically each worker uses one CPU core.

◇ Syntax

parpool(n) % n is the number of workers (a user-defined variable).

 % parallel codes

delete(gcp)

- ✓ On SCC, the default number of workers equals the number of CPU cores on the node.
- ✓ **gcp** (get current parpool) is a built-in variable.

parfor (1): Basics

- ◇ A simple implementation of parallel for-loop.
- ◇ Work load is distributed evenly and automatically based on loop index.
- ◇ Data starts on client (base workspace), automatically copy input data to workers' workspaces, and copy output data back to client when necessary calculation is done.
- ◇ Details are intentionally opaque to user. There are many additional restrictions as to what can and cannot be done in a **parfor** loop – this is the price of simplicity.

◇ Syntax

```
parfor i=1:n  
    % code block  
end
```

◇ An example:

```
x=zeros(1,12);  
parfor i=1:12  
    t = getCurrentTask(); disp(t.ID); % Display worker ID  
    x(i)=10*i; % Computation is done by workers simultaneously  
end
```

parfor (2): Rules for variables

- ◇ For the **parfor** loop to work, variables inside the loop must all fall into one of these categories:

Type	Description
Loop	A loop index variable for arrays
Sliced	An array whose segments are manipulated on different loop iterations
Broadcast	A variable defined before the loop and is used inside the loop but never modified
Reduction	Accumulates a value across loop iterations, regardless of iteration order
Temporary	Variable created inside the loop, but not used outside the loop

parfor (3): Modify variables

```
n=12; s = 0;
a = 100; b=50;
X = rand(1,n);
parfor k = 1 : n
    a = 2*k;      % k - loop index; a - temporary var: the value is not carried out of the loop
    Y(k) = X(k) + a*n;    % X, Y - sliced var; n - broadcast var
    b = 5;      % b - broadcast var:
    s = s + a;  % s - reduction var: : the value is carried out of the loop
end
```

- ◇ Question: what are the values of variables **a**, **b** and **s** after the parfor loop is done?

parfor (4): Reduction

- ◇ Reduction variables appear on both sides of an assignment statement, such as:

$X = X \text{ op } \text{expr}$

$X = \text{expr op } X$ (except subtraction)

- ✓ The operation *op* could be $+$, $-$, $*$, $.*$, $\&$, $|$

- ◇ A failed case: not a reduction:

```
x = 1;
parfor i= 1:10
    x = i - x;
end
```

- ◇ A successful case: a reduction:

```
x = 1;
parfor i= 1:10
    x = x - i;
end
```

parfor (5): Data dependency

◇ Data dependency: loop iterations must be independent

• A failed case:

```
n=10;  
a = 1:n;  
parfor i= 2:n  
    a(i) = a(i-1)*2;  
end  
% This may return unexpected results.
```

• A successful case:

```
n=10;  
a = 1:n;  
parfor i= 1:n  
    a(i) = a(i)*2;  
end  
% Each a(i) is read and modified by a  
worker. Different indexes are  
independent.
```

parfor (6): Loop index

◇ Loop index must be consecutive integers.

```
parfor i= 1 : 100    % OK
```

```
parfor i= -20 : 20   % OK
```

```
parfor i= 1 : 2 : 25 % No
```

```
parfor i= -7.5 : 7.5 % No
```

```
A = [3 7 -2 6 4 -4 9 3 7]; parfor i= find(A > 0) % No
```

parfor (7): Nested loops and functions

- ◇ The body of a **parfor**-loop
- ✓ can contain **for**-loops, including further nested **for**-loops.
- ✓ can not contain another **parfor**-loop.
- ✓ can make reference to a regular function but not a nested function.
- ✓ can call a function that contains another **parfor**-loop, which runs in parallel only if the outer **parfor**-loop runs serially (e.g. specifying one worker).

- ◇ Refer to: <https://www.mathworks.com/help/distcomp/nesting-and-flow-in-parfor-loops.html>

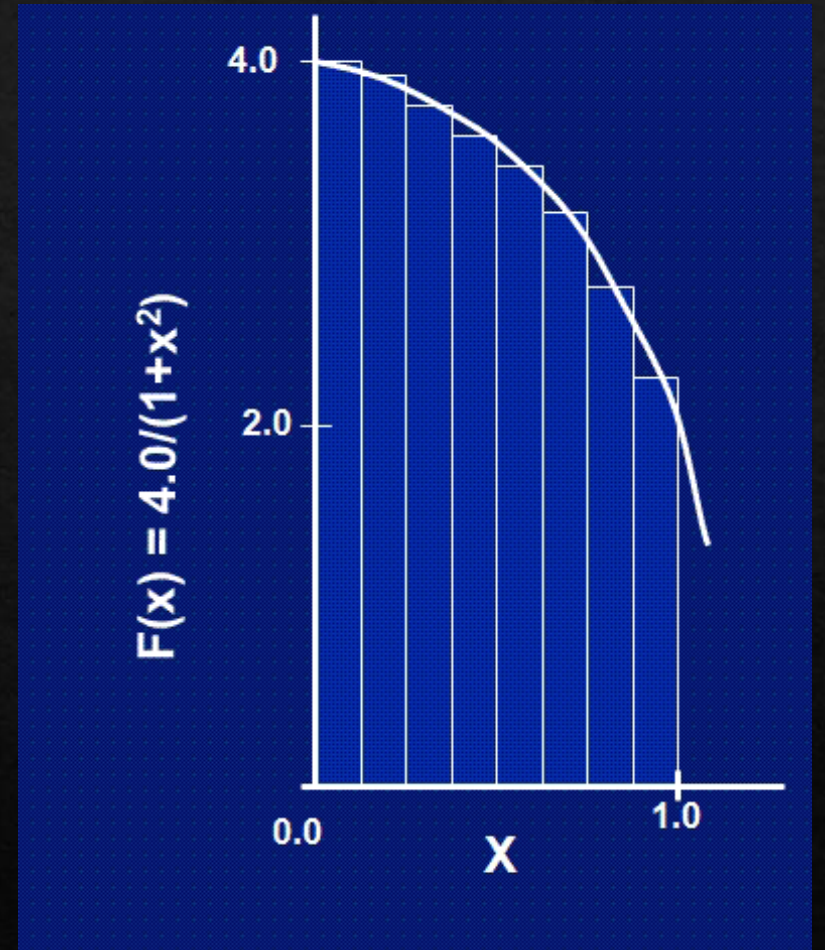
Compute the value of Pi

- ◇ Compute the value of Pi using the integral formula

$$\int_0^1 \frac{4.0}{(1+x^2)} dx = \pi$$

- ◇ The serial code

```
n=2000000000; dx=1/n; pi=0;
for i=1:n
    x = (i - 0.5) * dx;
    pi = pi + 4./(1.+x*x);
end
format long
pi=pi*dx
```



Exercise 1

◇ Compute the value of Pi using `parfor`

- i) Parallelize the code using `parfor` . Check whether all variables in the `parfor` region fall into one of the valid categories.
- ii) Compare the performances of the serial and the parallel codes.

spmd (1): Basics

◇ **spmd** = **S**ingle **P**rogram **M**ultiple **D**ata

Explicitly and/or automatically...

✓ divide work and data between workers/labs

✓ communicate between workers/labs

◇ **Syntax**

% execute on client/master out of spmd region

spmd

 % execute on all workers within spmd region

end

% execute on client/master out of spmd region

spmd (2): Number and index of workers

- Get an array chunk on each worker using built-in variables `numlabs` and `labindex`

```
parpool(4)
spmd
    disp(numlabs); % numlabs – total number of workers
    disp(labindex); % labindex – index of workers
    N=24;
    A=1:2:N;
    I = find(A > N*(labindex-1)/numlabs & A <= N*labindex/numlabs)
end
delete(gcf)
```

Pmode: Interactive Parallel Command Window

- ❑ Workers receive commands entered in the Parallel Command Window, process them, and send the command output back to the Parallel Command Window.

- ❑ Launch pmode

```
>> pmode start 4    % Request for 4 workers
```

- ❑ Execute commands in pmode (at prompt P>>)

```
P>> x = 2 * labindex  
P>> y = numlabs  
P>> if labindex == 1  
    z = x*10 + y  
end
```

spmd (3): Send and receive data

- ❑ `labSendReceive(ID_send_to, ID_receive_from, send_data)` - Send data to one worker and receive data from another worker.
- ❑ Example: circularly shift data between neighbor workers

```
spmd
```

```
  DataSent=labindex;
```

```
  right = mod(labindex, numlabs) + 1;      % the worker on the right
```

```
  left = mod(labindex - 2, numlabs) + 1;   % the worker on the left
```

```
  % Send data to the right and receive another data from left
```

```
  DataRcv = labSendReceive(right, left, DataSent)
```

```
end
```

spmd (4): Broadcast data

□ **labBroadcast** - Broadcast data from one worker to all other workers.

```
spmd
  source=1;
  if labindex == source
    data=1:12;
    % send data from the source worker to other workers, and save it in shared_data on the source worker.
    shared_data = labBroadcast(source, data)
  else
    % receive data on other workers and save it in share_data
    shared_data = labBroadcast(source)
  end
end
```

spmd (5): Composite variable and distributed array

- Use **Composite, distributed** out of **spmd** region

```
a=5; % Create a normal variable on client
b=Composite(); c=Composite(); % Create composite variables b and c on client
A=ones(4,4); A=distributed(A); % Create a matrix A on client and distribute it to workers
spmd
    x = a % Variable a is copied to workers and assigned to x. The local variable x is not accessible from client.
    y = labindex % Variable y is a local variable and is not accessible from client.
    b = labindex; % Composite variable b is modified by workers and is accessible from client
    c = magic(labindex+2); % Composite variable can be a matrix too.
    B = A * 2; % Computation is distributed to workers. The result matrix B is accessible from client.
end
b{:} % Output composite variable on client
c{:} % Output composite variable on client
B % Output distributed matrix on client
```

Distributed Matrix multiplication

- The `distributed` function can be used for parallel computing without using `spmd`.

```
A=randn(n); B=randn(n);  
a=zeros(n); b=zeros(n); c=zeros(n);  
  
parpool(4)  
a = distributed(A);    % Distributes A, B. a, b are distributed  
b = distributed(B);  
tic  
c = a * b;    % Run the multiplication in parallel by workers. c is distributed.  
toc  
delete(gcf)
```

spmd (6): Codistributed matrices

- Use `codistributed` within `spmd` region

```
n=1000; A = rand(n); B = rand(n); % create matrices A and B on client
spmd
    u = codistributed(A, codistributor1d(1)); % distribute A by row
    v = codistributed(B, codistributor1d(2)); % distribute B by column, so that A and B are codistributed.
    w = u * v; % run in parallel by workers; the result w is distributed.
    p = rand(n, codistributor1d(1)); % create distributed matrix p on workers
    q = codistributed.rand(n); % create distributed matrix q on workers; p and q are codistributed
    s = p * q; % run in parallel by workers; the result s is distributed
end
x=3+w % use w directly on client
y=2*s % use s directly on client
```

Exercise 2

◇ Compute the value of Pi (using `spmd`)

- i) Write a parallel code for computing the value of Pi using `spmd` .
- ii) Compare the performances of the serial, the `parfor` parallel and the `spmd` parallel codes.

(Hints: Distribute the grid to workers and compute local sum on all workers, then use the function `gplus` to compute the total sum.)

A Solution to Exercise 2

```
n=500000000; dx=1/n; total_sum=Composite(); % total_sum will be modified in and used out of spmd region
tic % start measuring time
spmd % start spmd region
    m=n/numlabs; % number of grid points on each worker
    length=1/numlabs; % grid length on each worker
    startx = (labindex -1)*length; % starting x of the current worker
    endx = labindex*length; % ending x of the current worker
    x = startx : dx : endx; % the portion of x held by the current worker
    local_sum=0; % set 0 before accumulating
    local_sum = sum( 4. / (1. + x .* x) ); % compute local sum on the current lab
    total_sum = gplus(local_sum, 1); % add up all local sums and store it on lab 1
end % end spmd region
toc % end measuring time
format long
pi=total_sum{1}*dx % get the value of total_sum from worker 1 and output the result on client
```

Using Matlab on GPU (1)

- ◇ For many problems, GPUs achieve better performance than CPUs.
- ◇ MATLAB GPU utilities are growing.
- ◇ Matrix operations on GPU:

```
n = 6400;           % matrix size, better to be multiple of GPU warp-size (i.e. 32).
a = rand(n);       % create n * n random matrix a on base workspace (host)
A = gpuArray(a);   % A is created on GPU. The value of a is copied to A.
B = gpuArray.rand(n); % Create random matrix directly on GPU
C = A * B;         % Matrix multiplication is computed on GPU
c = gather(C);     % bring result back to base workspace on CPU/host
```

Using Matlab on GPU (2)

- ◇ Run built-In functions on a GPU.
- ◇ <https://www.mathworks.com/help/distcomp/run-built-in-functions-on-a-gpu.html>
- ◇ Examples: Run fast-fourier-transform on GPU:

```
Ga = rand(1000,'single','gpuArray');  
Gfft = fft(Ga);  
Gb = (real(Gfft) + Ga) * 6;  
G = gather(Gb);  
whos
```

Using Matlab on GPU (3)

◆ **arrayfun**: Apply function to each element of array on GPU.

```
n=10;  
a = rand(n,1,'gpuArray');    % create random arrays on GPU  
b = rand(n,1,'gpuArray');  
c = rand(n,1,'gpuArray');  
R = arrayfun( @(x,y,z)(x.*y+z), a, b, c );    % compute arrayfun on GPU  
results = gather(R)    % bring result back to base workspace on CPU/host
```

Using Matlab on GPU (4)

- ◇ Use CUDA functions in Matlab.
- ◇ A CUDA C function for adding two vectors.

```
__global__ void add2( double * v1, const double * v2 ) {  
    int idx = threadIdx.x;  
    v1[idx] += v2[idx];  
}
```

- ◇ Compile the CUDA code to get an assembly-level ptx file

```
$ module load cuda/9.1
```

```
$ nvcc -ptx vecadd.cu
```

Using Matlab on GPU (5)

◇ A Matlab code to call the CUDA function.

```
% Create GPU CUDA kernel object from PTX and CU code
k = parallel.gpu.CUDAKernel('addvec.ptx','addvec.cu','add2');
N = 128;
k.ThreadBlockSize = N;      % Array size: better to be multiple of 32.
in1 = ones(N,1,'gpuArray'); % A GPU array with all elements equal to one.
in2 = rand(N,1,'gpuArray'); % A GPU array with random values between 0 and 1.
result = feval(k,in1,in2);  % Run the kernel function on GPU and return the result.
disp(result)                % Display the result.
```

Further Information

- ◇ **MathWorks Web:**

- ✓ MATLAB Parallel Computing Toolbox documentation:

<http://www.mathworks.com/help/distcomp/index.html>

- ◇ **BU Research Computing Services (RCS) Web:**

- ✓ MATLAB Parallel Computing Toolbox:

<http://www.bu.edu/tech/support/research/software-and-programming/common-languages/matlab/pct/>

- ◇ **A book:** *Accelerating MATLAB Performance: 1001 tips to speed up MATLAB programs* by Yair M. Altman

- ◇ **RCS help:** help@scc.bu.edu , shaohao@bu.edu