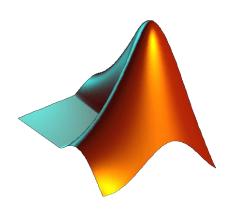
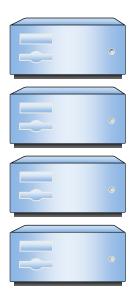


Parallel Computing with MATLAB





Tim Mathieu Sr. Account Manager

Gerardo Hernandez

Application Engineer

Abhishek Gupta

Application Engineer



Some Questions to Consider

- Do you want to speed up your algorithms?
- Do you have datasets too big to fit on your computer?

If so...

- Do you have a multi-core or multiprocessor desktop machine?
- Do you have access to a computer cluster?



Solving Big Technical Problems

Challenges You could... Solutions Long running Larger Compute Pool Wait (e.g. More Processors) Computationally intensive Reduce size Larger Memory Pool Large data set of problem (e.g. More Machines)

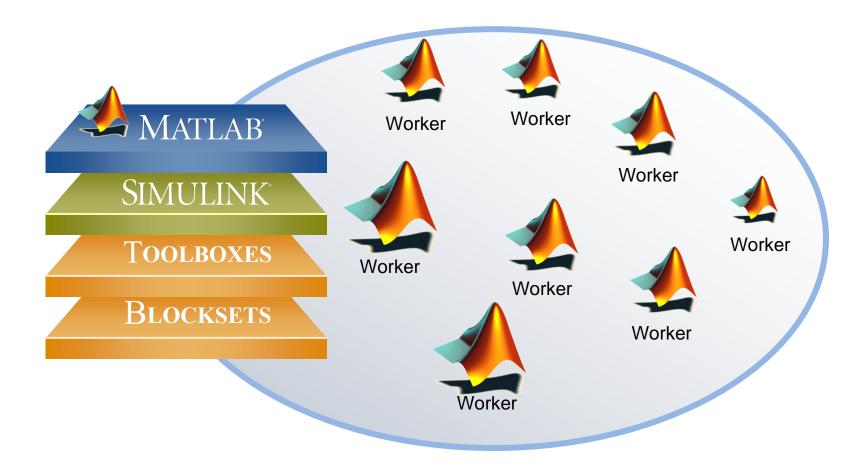


Utilizing Additional Processing Power

- Built-in multithreading
 - Core MATLAB
 - Introduced in R2007a
 - Utility for specific matrix operations
 - Automatically enabled since R2008a
- Parallel computing tools
 - Parallel Computing Toolbox
 - MATLAB Distributed Computing Server
 - Broad utility controlled by the MATLAB user

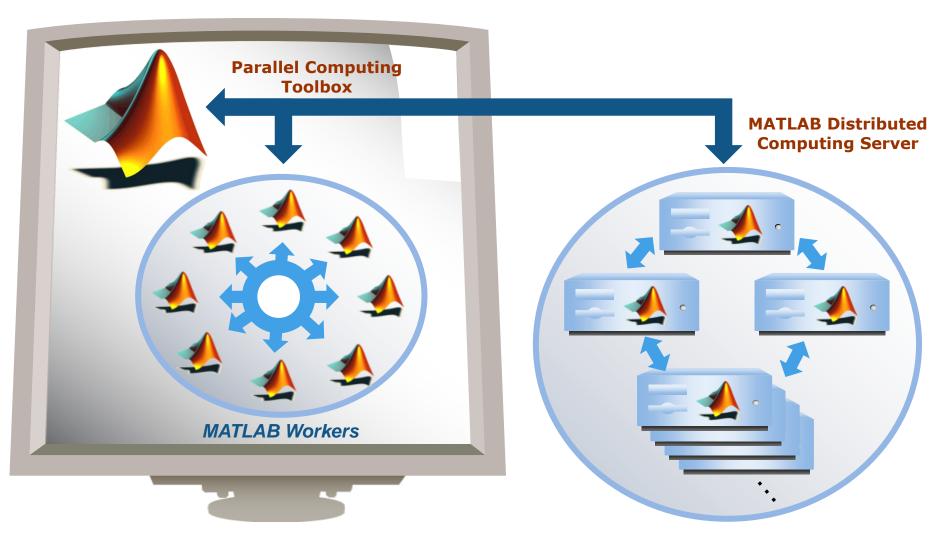


Parallel Computing with MATLAB





Parallel Computing with MATLAB



User's Desktop

Compute Cluster



Programming Parallel Applications

Level of control

Required effort

Minimal

None

Some

Straightforward

Extensive

Involved



Programming Parallel Applications

Level of control

Minimal

Some

Extensive

Parallel Options

Support built into Toolboxes

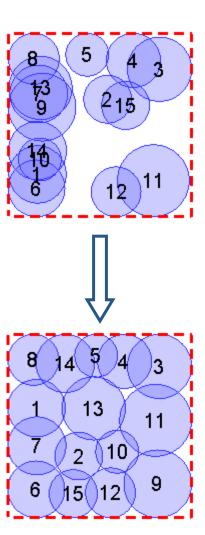
High-Level
Programming Constructs:
(e.g. parfor, batch, distributed)

Low-Level
Programming Constructs:
(e.g. Jobs/Tasks, MPI-based)



Example: Optimizing Tower Placement

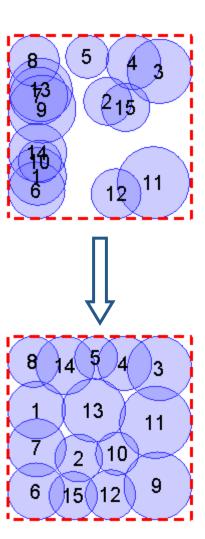
- Determine location of cell towers
- Maximize coverage
- Minimize overlap





Summary of Example

- Enabled built-in support for Parallel Computing Toolbox in Optimization Toolbox
- Used a pool of MATLAB workers
- Optimized in parallel using fmincon





Parallel Support in Optimization Toolbox

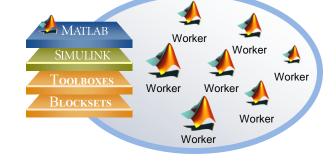
• Functions:

- fmincon
 - Finds a constrained minimum of a function of several variables
- fminimax
 - Finds a minimax solution of a function of several variables
- fgoalattain
 - Solves the multiobjective goal attainment optimization problem
- Functions can take finite differences in parallel in order to speed the estimation of gradients



Tools with Built-in Support

- Optimization Toolbox
- Global Optimization Toolbox
- Statistics Toolbox
- SystemTest
- Simulink Design Optimization
- Bioinformatics Toolbox
- Model-Based Calibration Toolbox



• . . .

http://www.mathworks.com/products/parallel-computing/builtin-parallel-support.html

Directly leverage functions in Parallel Computing Toolbox



Programming Parallel Applications

Level of control

Minimal

Some

Extensive

Parallel Options

Support built into Toolboxes

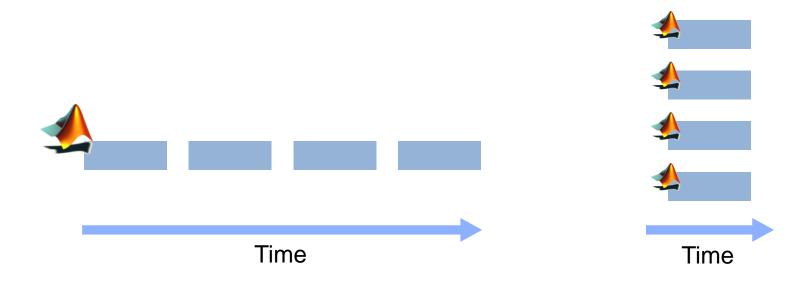
High-Level
Programming Constructs:
(e.g. parfor, batch, distributed)

Low-Level
Programming Constructs:
(e.g. Jobs/Tasks, MPI-based)



Running Independent Tasks or Iterations

- Ideal problem for parallel computing
- No dependencies or communications between tasks
- Examples include parameter sweeps and Monte Carlo simulations



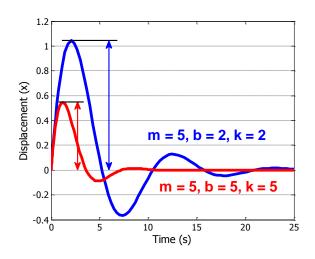


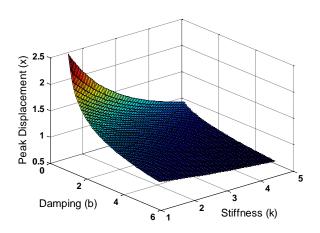
Example: Parameter Sweep of ODEs

Solve a 2nd order ODE

$$m\ddot{x} + b_{1,2,...} \dot{x} + k_{2,...} x = 0$$

- Simulate with different values for b and k
- Record peak value for each run
- Plot results

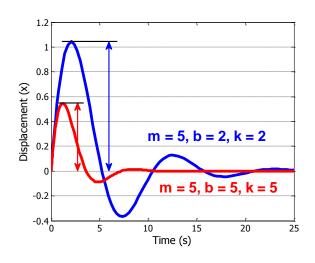


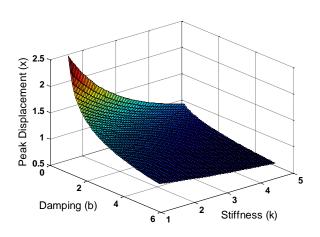




Summary of Example

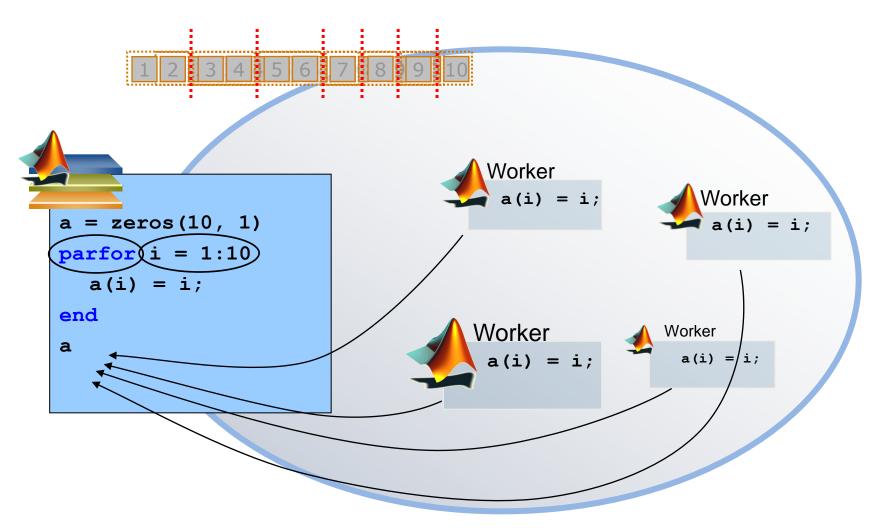
- Mixed task-parallel and serial code in the same function
- Ran loops on a pool of MATLAB resources
- Used Code Analyzer to help in converting existing for-loop into parfor-loop







The Mechanics of parfor Loops



Pool of MATLAB Workers



Converting for to parfor

- Requirements for parfor loops
 - Task independent
 - Order independent
- Constraints on the loop body
 - Cannot "introduce" variables (e.g. eval, load, global, etc.)
 - Cannot contain break or return statements
 - Cannot contain another parfor loop



Advice for Converting for to parfor

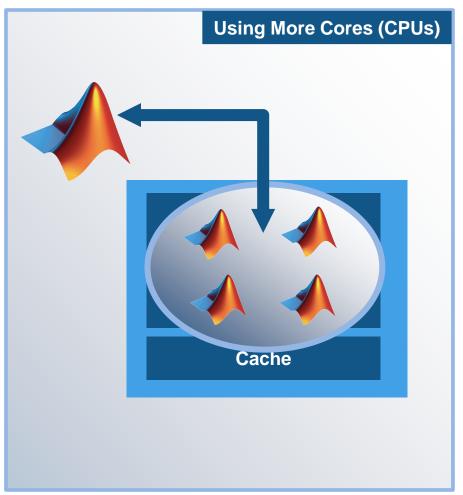
Use Code Analyzer to diagnose parfor issues

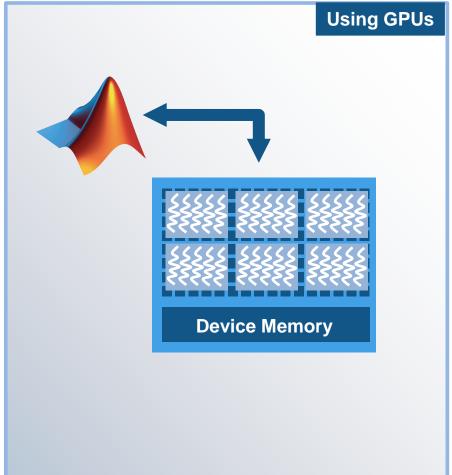
 If your for loop cannot be converted to a parfor, consider wrapping a subset of the body to a function

- Read the section in the documentation on classification of variables
- http://blogs.mathworks.com/loren/2009/10/02/using-parforloops-getting-up-and-running/



Performance Gain with More Hardware







What is a Graphics Processing Unit (GPU)

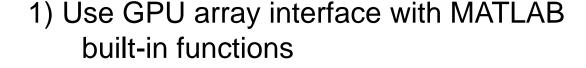
- Originally for graphics acceleration, now also used for scientific calculations
- Massively parallel array of integer and floating point processors
 - Typically hundreds of processors per card
 - GPU cores complement CPU cores
- Dedicated high-speed memory





^{*} Parallel Computing Toolbox requires NVIDIA GPUs with Compute Capability 1.3 or greater, including NVIDIA Tesla 10-series and 20-series products. See http://www.nvidia.com/object/cuda_gpus.html for a complete listing

Summary of Options for Targeting GPUs

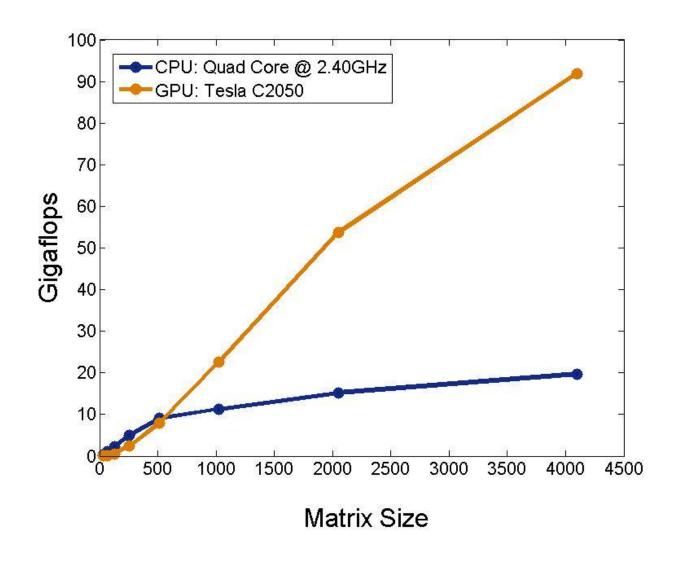


2) Execute custom functions on elements of the GPU array

 Invoke your CUDA kernels directly from MATLAB

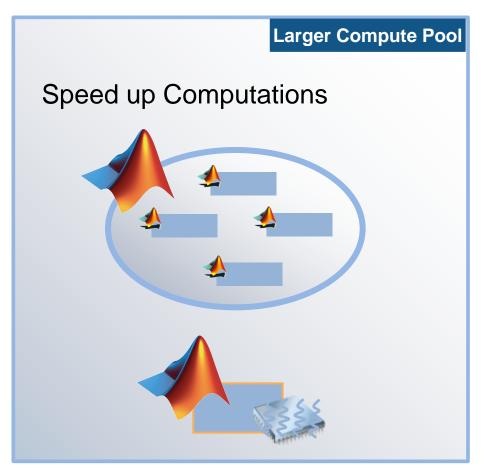


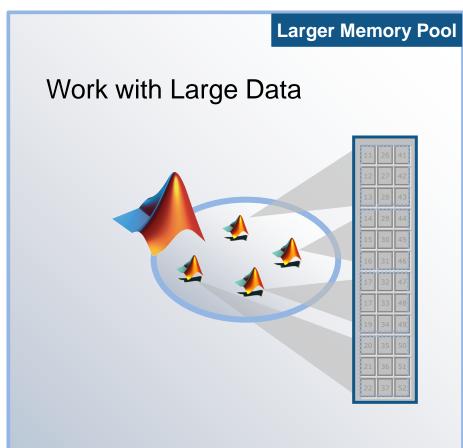
Performance: A\b with Double Precision





Parallel Computing enables you to ...





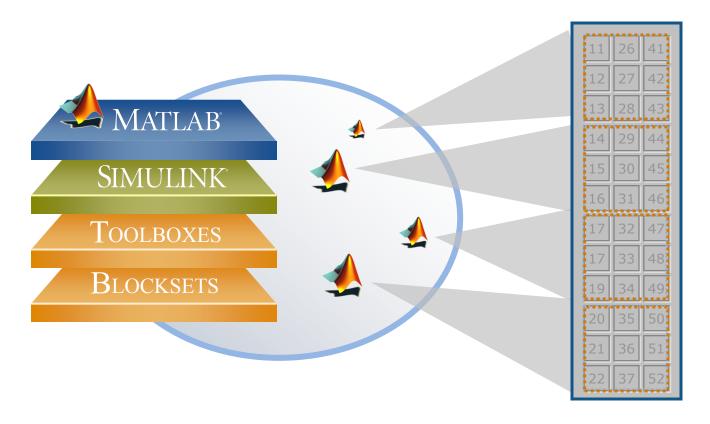


Limited Process Memory

- 32-bit platforms
 - Windows 2000 and XP (by default): 2 GB
 - Linux/UNIX/MAC system configurable: 3-4 GB
 - Windows XP with /3gb boot.ini switch: 3 GB
- 64-bit platforms
 - Linux/UNIX/MAC: 8 TB
 - Windows XP Professional x64: 8TB



Client-side Distributed Arrays



Remotely Manipulate Array from Desktop

Distributed Array Lives on the Cluster



Enhanced MATLAB Functions That Operate on Distributed Arrays

| Type of Function | Function Names |
|--|--|
| Data functions | cumprod, cumsum, fft, max, min, prod, sum |
| Data type functions | <pre>cast, cell2mat, cell2struct, celldisp, cellfun, char, double, fieldnames, int16, int32, int64, int8, logical, num2cell, rmfield, single, struct2cell, swapbytes, typecast, uint16, uint32, uint64, uint8</pre> |
| Elementary and trigonometric functions | abs, acos, acosd, acosh, acot, acotd, acoth, acsc, acscd, acsch, angle, asec, asecd, asech, asin, asind, asinh, atan, atan2, atand, atanh, ceil, complex, conj, cos, cosd, cosh, cot, cotd, coth, csc, cscd, csch, exp, expm1, fix, floor, hypot, imag, isreal, log, log10, log1p, log2, mod, nextpow2, nthroot, pow2, real, reallog, realpow, realsqrt, rem, round, sec, secd, sech, sign, sin, sind, sinh, sqrt, tan, tand, tanh |
| Elementary matrices | cat, diag, eps, find, isempty, isequal, isequalwithequalnans, isfinite, isinf, isnan, length, ndims, size, tril, triu |
| Matrix functions | chol, eig, lu, norm, normest, svd |
| Array operations | all, and, any, bitand, bitor, bitxor, ctranspose, end, eq, ge, gt, horzcat, ldivide, le, lt, minus, mldivide, mrdivide, mtimes, ne, not, or, plus, power, rdivide, subsasgn, subsindex, subsref, times, transpose, uminus, uplus, vertcat, xor |
| Sparse matrix functions | full, issparse, nnz, nonzeros, nzmax, sparse, spfun, spones |
| Special functions | <u>dot</u> |



spmd blocks

end

```
spmd
% single program across workers
```

- Mix parallel and serial code in the same function
- Run on a pool of MATLAB resources
- Single Program runs simultaneously across workers
 - Distributed arrays, message-passing
- Multiple Data spread across multiple workers
 - Data stays on workers



Programming Parallel Applications

Level of control

Minimal

Some

Extensive

Parallel Options

Support built into Toolboxes

High-Level
Programming Constructs:
(e.g. parfor, batch, distributed)

Low-Level
Programming Constructs:
(e.g. Jobs/Tasks, MPI-based)



MPI-Based Functions in Parallel Computing Toolbox™

Use when a high degree of control over parallel algorithm is required

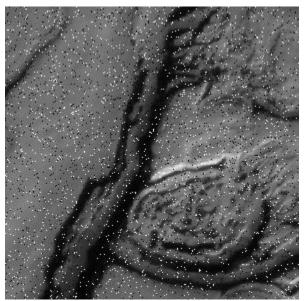
- High-level abstractions of MPI functions
 - labSendReceive, labBroadcast, and others
 - Send, receive, and broadcast any data type in MATLAB
- Automatic bookkeeping
 - Setup: communication, ranks, etc.
 - Error detection: deadlocks and miscommunications
- Pluggable
 - Use any MPI implementation that is binary-compatible with MPICH2



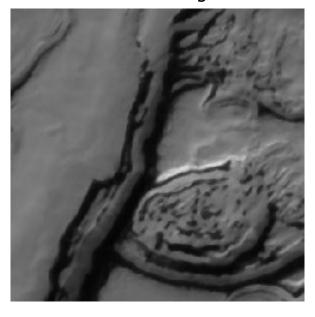
Example: Image De-Noising (Large Image Processing)

Use median filtering to reduce "salt & pepper" noise.

Noisy Image



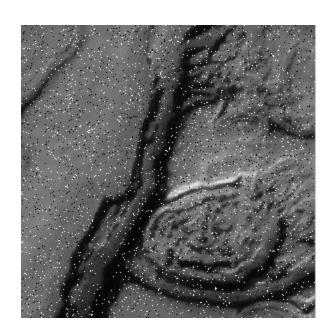
Filtered Image



http://hirise.lpl.arizona.edu/ From - NASA/JPL/University of Arizona

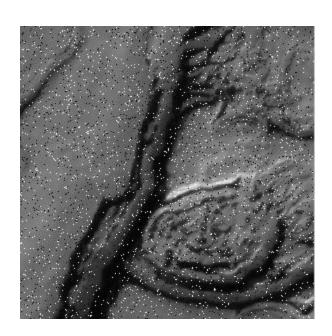


Noisy Image – too large for a desktop



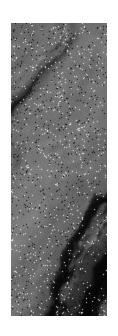


Distribute Data

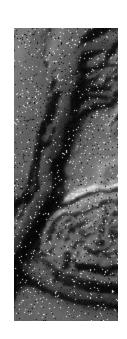




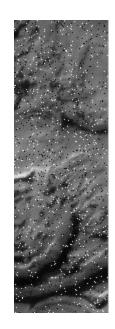
Distribute Data







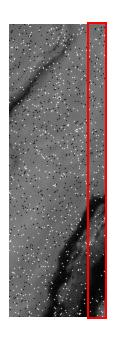




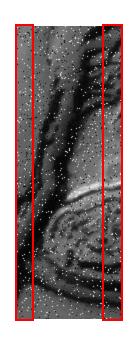




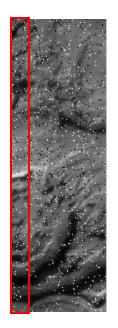
Pass Overlap Data







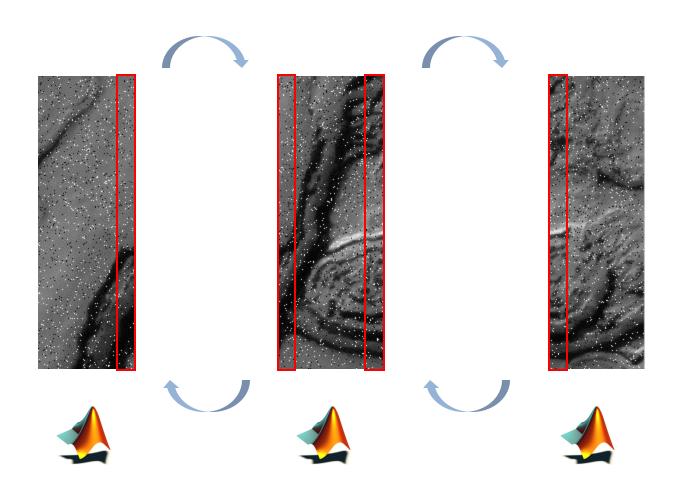






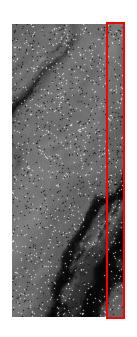


Pass Overlap Data

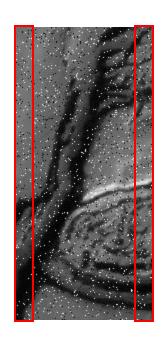




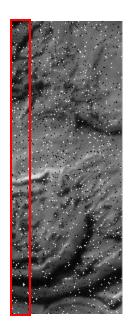
Pass Overlap Data







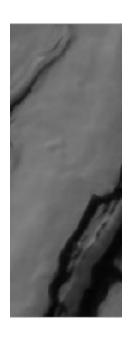




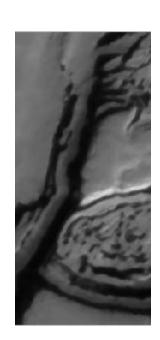




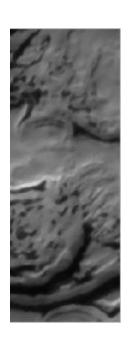
Apply Median Filter







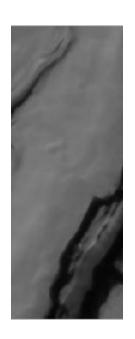


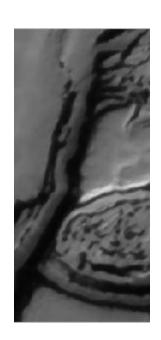


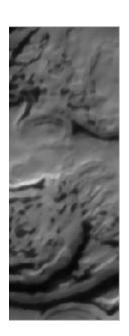




Combine as Distributed Data

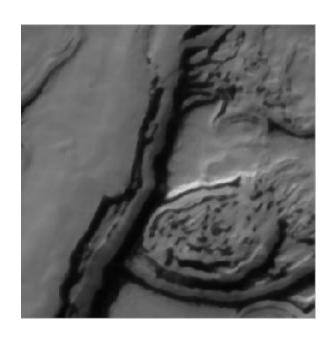






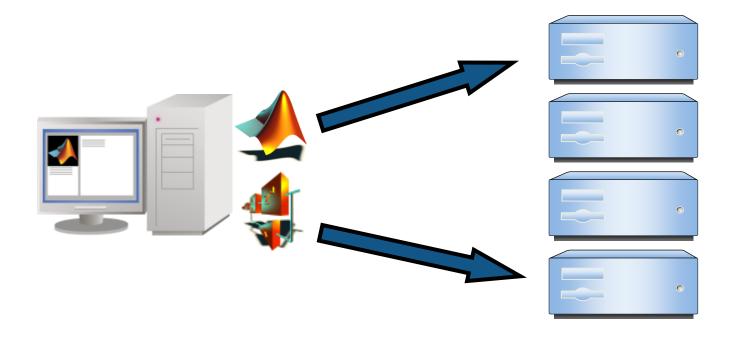


Combine as Distributed Data





Scheduling Applications





Interactive to Scheduling

Interactive

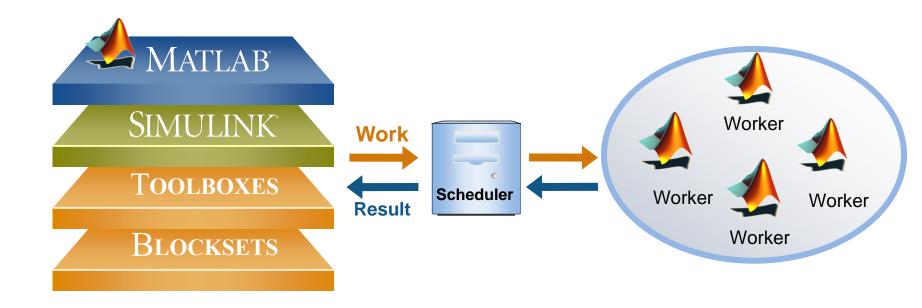
- Great for prototyping
- Immediate access to MATLAB workers

Scheduling

- Offloads work to other MATLAB workers (local or on a cluster)
- Access to more computing resources for improved performance
- Frees up local MATLAB session



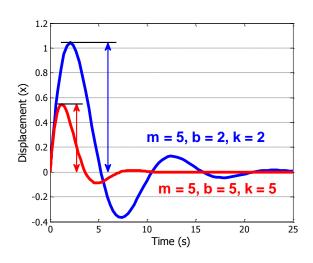
Scheduling Work

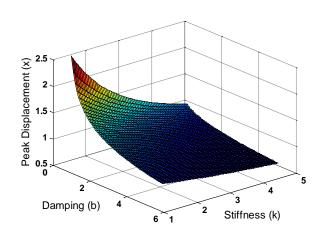




Example: Schedule Processing

- Offload parameter sweep to local workers
- Get peak value results when processing is complete
- Plot results in local MATLAB

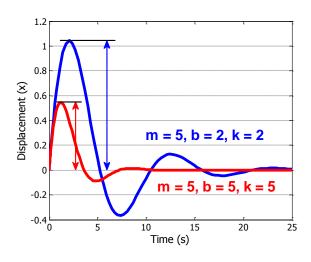


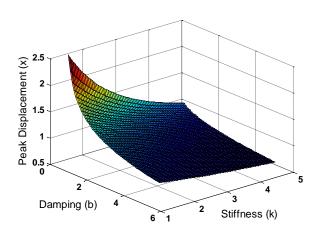




Summary of Example

- Used batch for off-loading work
- Used matlabpool option to off-load and run in parallel
- Used load to retrieve worker's workspace







Scheduling Workflows

parfor

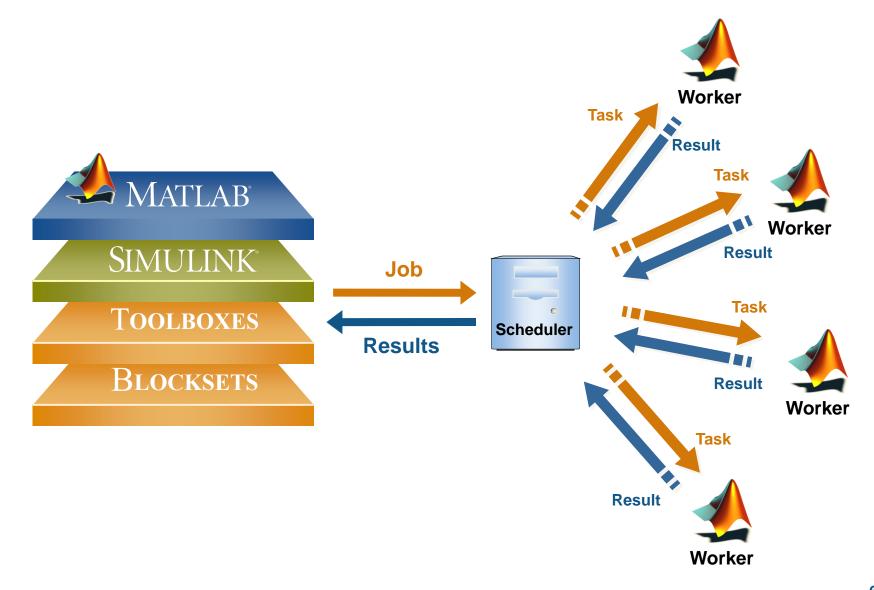
- Multiple independent iterations
- Easy to combine serial and parallel code
- Workflow
 - Interactive using matlabpool
 - Scheduled using batch

jobs/tasks

- Series of independent tasks; not necessarily iterations
- Workflow → Always scheduled



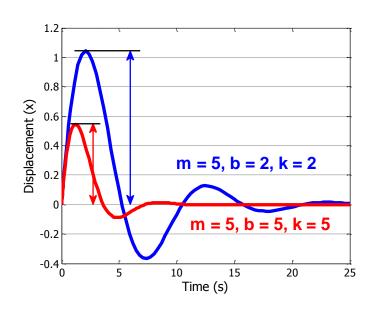
Scheduling Jobs and Tasks





Example: Scheduling Independent Simulations

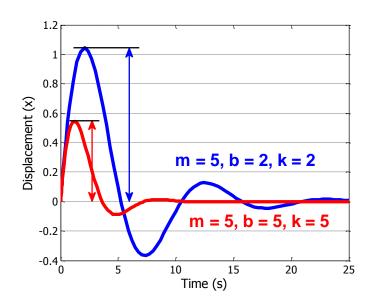
- Offload three independent approaches to solving our previous ODE example
- Retrieve simulated displacement as a function of time for each simulation
- Plot comparison of results in local MATLAB





Summary of Example

- Used findResource to find scheduler
- Used createJob and createTask to set up the problem
- Used submit to off-load and run in parallel
- Used getAllOutputArguments to retrieve all task outputs





Factors to Consider for Scheduling

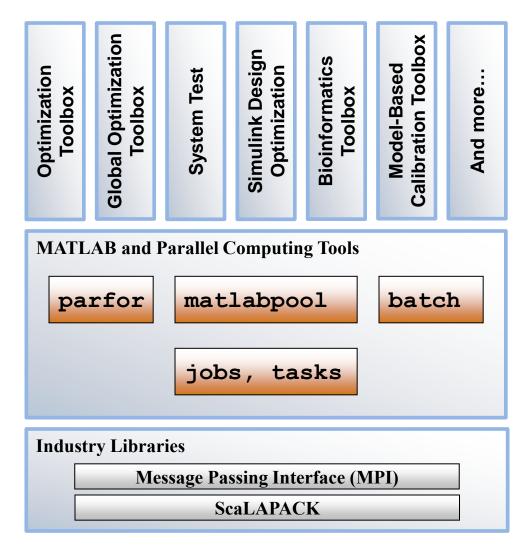
- There is always an overhead to distribution
 - Combine small repetitive function calls
- Share code and data with workers efficiently
 - Set job properties (FileDependencies, PathDependencies)
- Minimize I/O
 - Enable Workspace option for batch
- Capture command window output
 - Enable CaptureDiary option for batch



Parallel Computing with MATLAB

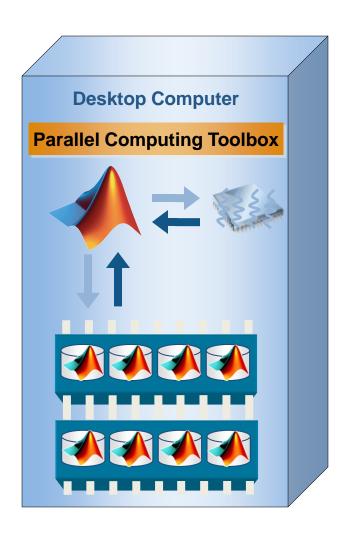
 Built in parallel functionality within specific toolboxes (also requires Parallel Computing Toolbox)

- High level parallel functions
- Low level parallel functions
- Built on industry standard libraries





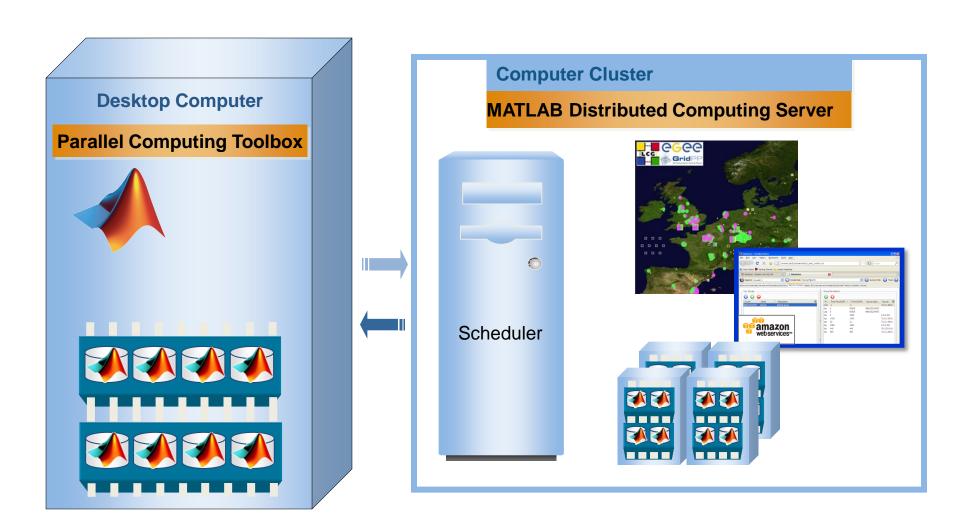
Parallel Computing on the Desktop



- Rapidly develop parallel applications on local computer
- Take full advantage of desktop power by using CPUs and GPUs
- Separate computer cluster not required



Scale Up to Clusters, Grids and Clouds





Licensing: MATLAB[®] Distributed Computing Server[™]

MATLAB

Simulink

Toolboxes

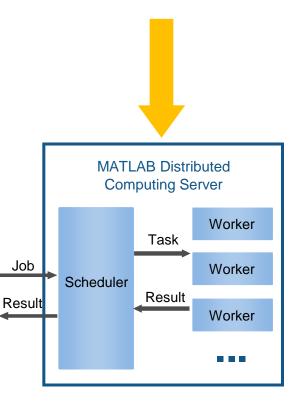
Blocksets

Parallel

Computing

Toolbox

- One key required per worker:
 - Packs of 8, 16, 32, 64, 128, etc.
 - Worker is a MATLAB® session, not a processor
- All-product install
 - No code generation or deployment products





Support for Schedulers

Direct Support













Open API for others







MathWorks Contact Information

For pricing, licensing, trials and general questions:

Tim Mathieu

Sr. Account Manager

Education Sales Department

Email: <u>Tim.Mathieu@mathworks.com</u>

Phone: 508.647.7016

Customer Service: service@mathworks.com

508.647.7000

Technical Support: support@mathworks.com

508.647.7000