MATLAB EXPO 2017

Parallel Computing with MATLAB and Simulink

Alka Nair Application Engineer MathWorks India Private Limited



Why Parallel Computing ?

- Size and complexity of analytical problems is growing across industries
- Need faster insight to bring competitive products to market quickly
- Hardware is becoming powerful: Leverage computational power of multicore desktops, GPUs, clusters

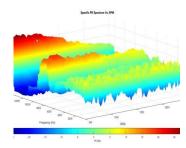


Key Takeaways

- > Overcome challenges with MathWorks Parallel Computing Tools
 - Save engineering and research time and focus on results
 - Leverage computational power of broadly available hardware with minimal changes to your existing code
 - (Multicore Desktops, GPUs, Clusters)
 - Seamlessly scale from your desktop to clusters or the cloud
 - Speed-up analysis of Big Data using built-in parallel computing capabilities



Where is Parallel Computing Used?



Bosch Develops Platform for Automotive Test Data Analysis and Visualization Validation time reduced by 40-50% **3-4 months of development time saved**

Heart Transplant Studies 4 weeks reduced to 5 days





Carnegie Wave Energy Designs and Builds Wave Energy Farm Sensitivity studies accelerated 12x

Lockheed Martin Builds Discrete-Event Model of Fleet Performance Simulation time reduced from months to hours 20X faster simulation time Linkage with Neural Network Toolbox



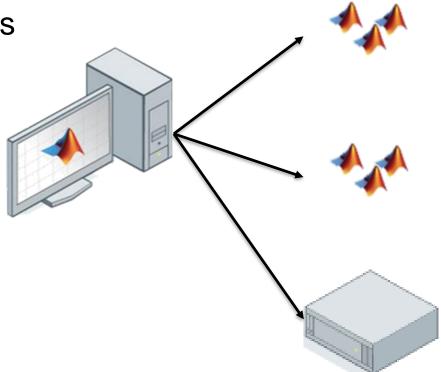


Commerzbank Develops Production Software Systems for Calculating Derived Market Data Implementation time reduced by months Updates loaded 8X faster



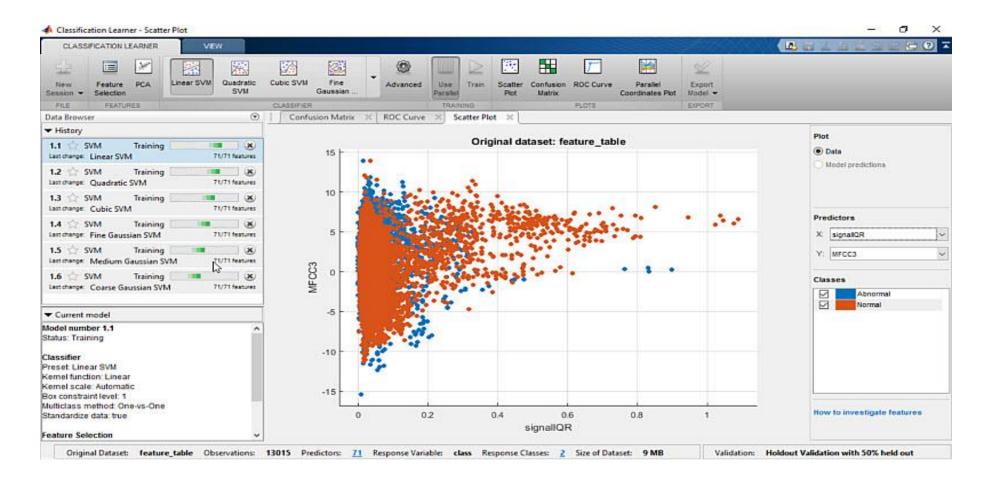
Agenda

- Parallel computing paradigms in MATLAB and Simulink
- Accelerate applications with NVIDIA GPUs
- Scaling to clusters and clouds
- Summary



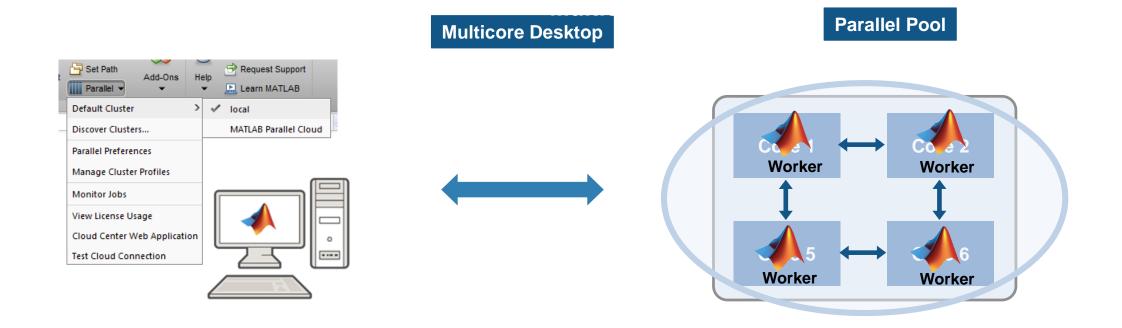


Classification learner demo



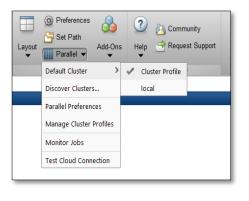


Parallel Computing Paradigm - Hardware Multicore Desktops

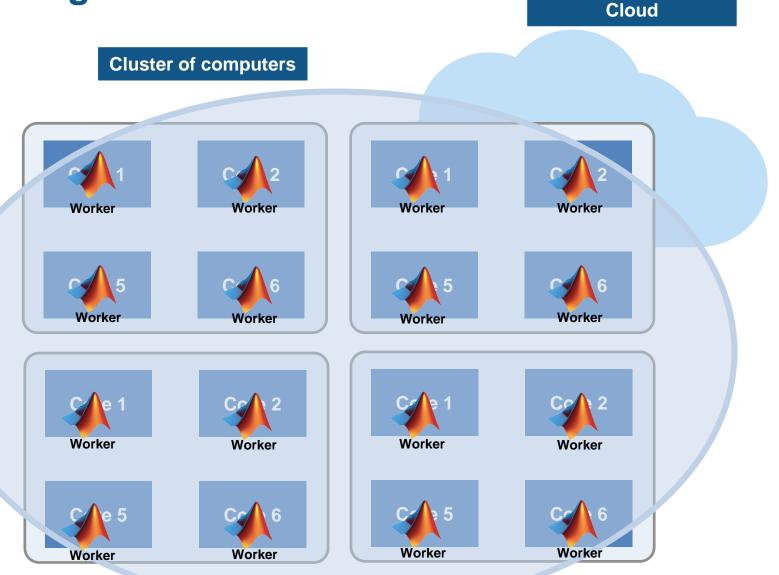




Parallel Computing Paradigm - Hardware Cluster Hardware









Programming Parallel Applications – Programming Constructs



Parallel-enabled toolboxes

Simple programming constructs

Advanced programming constructs



Parallel Computing: Neural Network Toolbox

%% Load data set
[x, t] = bodyfat_dataset;
%% Define the network
net1 = feedforwardnet(10);
%% Use parallel Computing to train the Network
net2 = train(net1,x,t,'useParallel','yes');
y = net2(x,'useParallel','yes');



Parallel-enabled Toolboxes (MATLAB® Product Family)

Enable parallel computing support by setting a flag or preference

Image Processing

Batch Image Processor, Block Processing, GPU-enabled functions

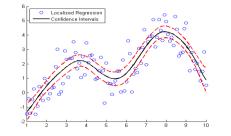




Original Image of Peppers Recolored Image of Peppers

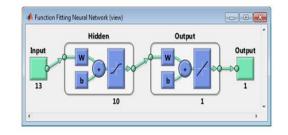
Statistics and Machine Learning

Resampling Methods, k-Means clustering, GPU-enabled functions



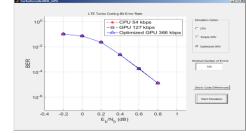
Neural Networks

Deep Learning, Neural Network training and simulation



Signal Processing and Communications

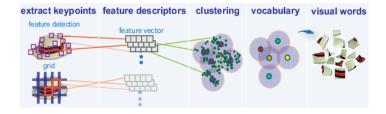
GPU-enabled FFT filtering, cross correlation, BER



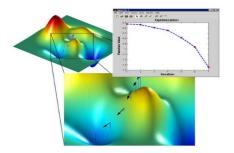
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Computer Vision

Parallel-enabled functions in bag-of-words workflow



Optimization Parallel estimation of gradients



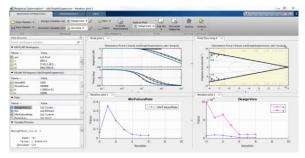


Parallel-enabled Toolboxes (Simulink® Product Family)

Enable parallel computing support by setting a flag or preference

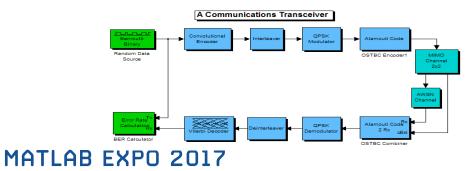
Simulink Design Optimization

Response optimization, sensitivity analysis, parameter estimation



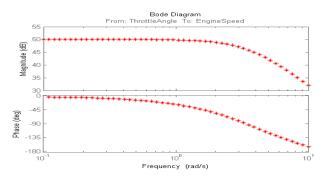
Communication Systems Toolbox

GPU-based System objects for Simulation Acceleration



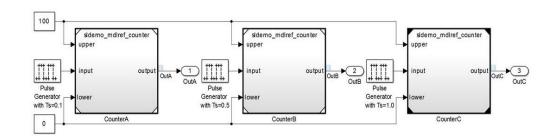
Simulink Control Design

Frequency response estimation



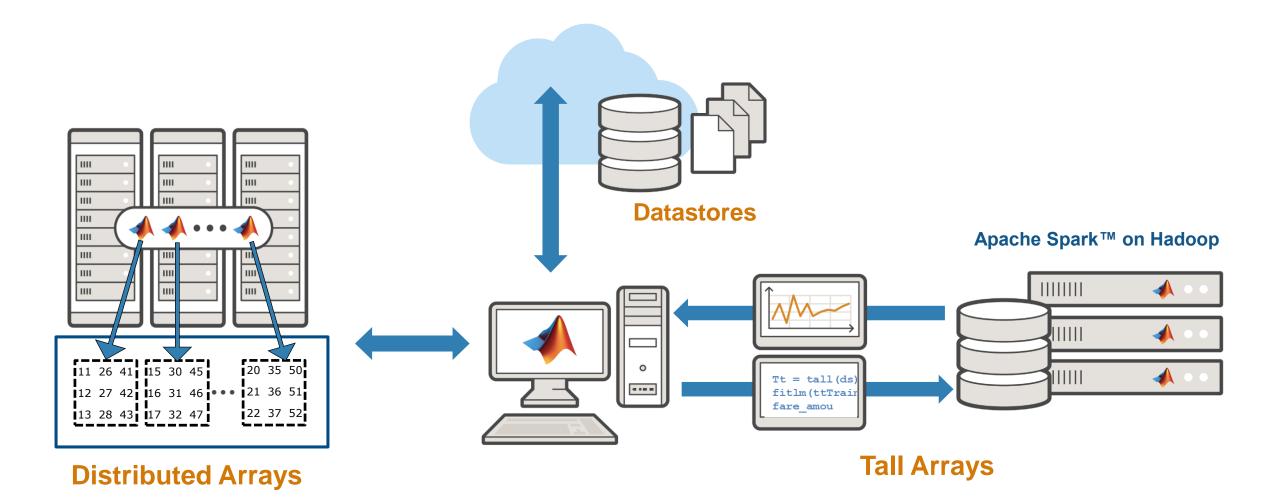
Simulink/Embedded Coder

Generating and building code



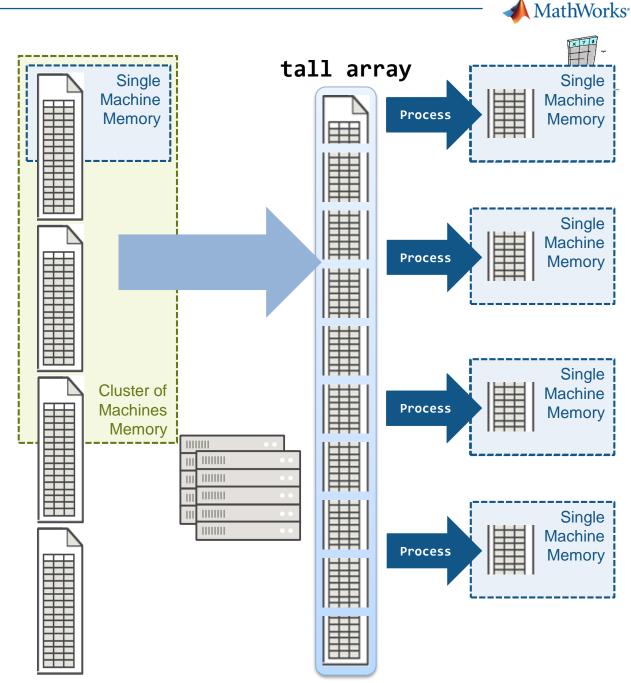


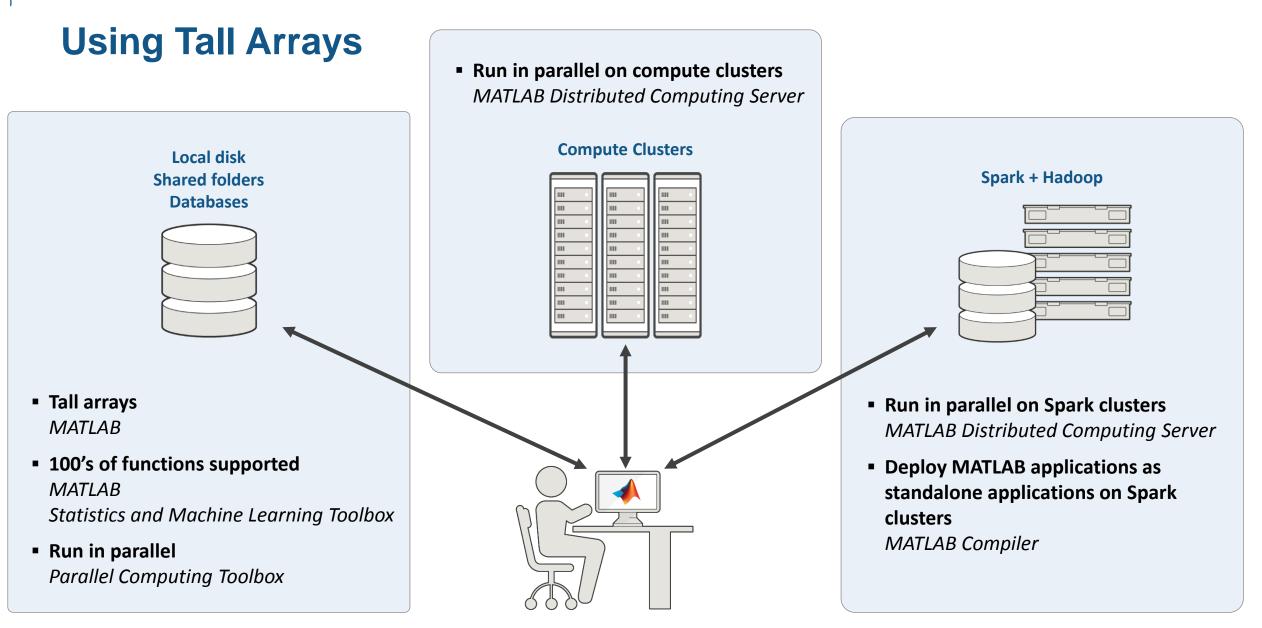
Data Intensive: Big Data support in MATLAB



tall arrays R2016b

- With Parallel Computing Toolbox, process several "chunks" at once
- Can scale up to clusters with MATLAB Distributed Computing Server







Programming Parallel Applications – Programming Constructs

Ease of Use

Parallel-enabled toolboxes

Simple programming constructs Eg. Parfor, Batch

Advanced programming constructs Spmd,createJob,labsend



Demo: Getting Data from a Web API

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<pre>or idx = 1:numPhotos info = photoInfo{idx}; [photo, username] = getPhotoFromFlickrPhotoInfo(appKey, info); % Plot each one and display attributions</pre>		
<pre>ior idx = 1:numPhotos info = photoInfo{idx}; [photo, username] = getPhotoFromFlickrPhotoInfo(appKey, info); % Plot each one and display attributions plotDataAndAttribution(axesHandles(idx), photo, info, username, idx);</pre>	s) ;	
<pre>for idx = 1:numPhotos info = photoInfo{idx}; [photo, username] = getPhotoFromFlickrPhotoInfo(appKey, info); % Plot each one and display attributions plotDataAndAttribution(axesHandles(idx), photo, info, username, idx); drawnow();</pre>);	
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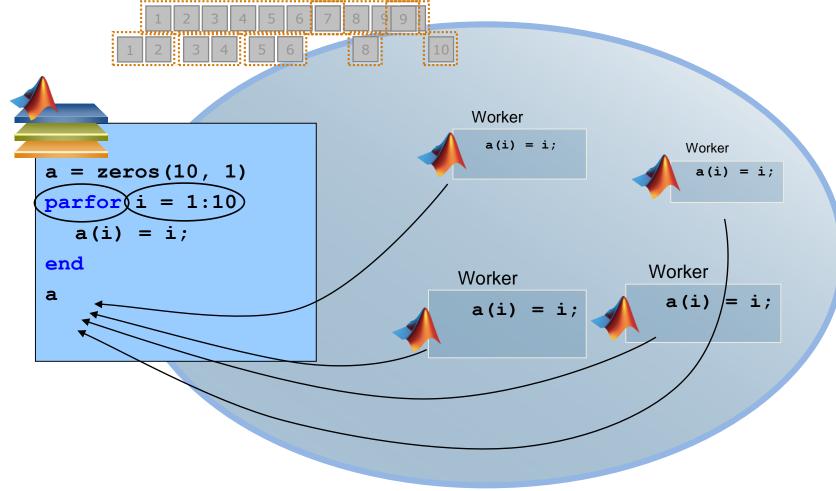


Demo: Getting Data from a Web API using parfor

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<pre>numPhotos = 30; tag = 'silverstone'; photoInfo = searchFlickr(appKey, tag, numPhotos);</pre>	
Download the photos and display them	
pre-allocate results	
<pre>photos = cell(numPhotos, 1); usernames = cell(numPhotos, 1); % Then retrieve the photos in a parfor loop. tic; parfor idx = 1:numPhotos [photos{idx}, usernames{idx}] = getPhotoFromFlickrPhotoInfo(appKey, photoInf end t = toc; fprintf('Retrieving %d photos using parfor loop took %0.3f seconds\n', numPhotos</pre>	
<pre>numPlotsAcross = 6; [figHandle, axesHandles] = createFigure(numPhotos, numPlotsAcross); figure(figHandle); for idx = 1:numPhotos plotDataAndAttribution(axesHandles(idx), photos{idx}, photoInfo{idx}, usernames{idx}, idx);</pre>	
end	



Explicit Parallelism: Independent Tasks or Iterations Simple programming constructs using parfor, parfeval Examples: parameter sweeps, Monte Carlo simulations No dependencies or communications between tasks

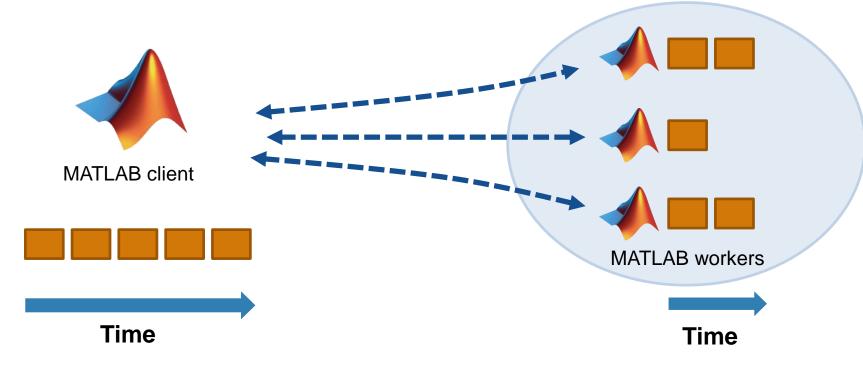




Independent Tasks or Iterations

Simple programming constructs using parfor, parfeval

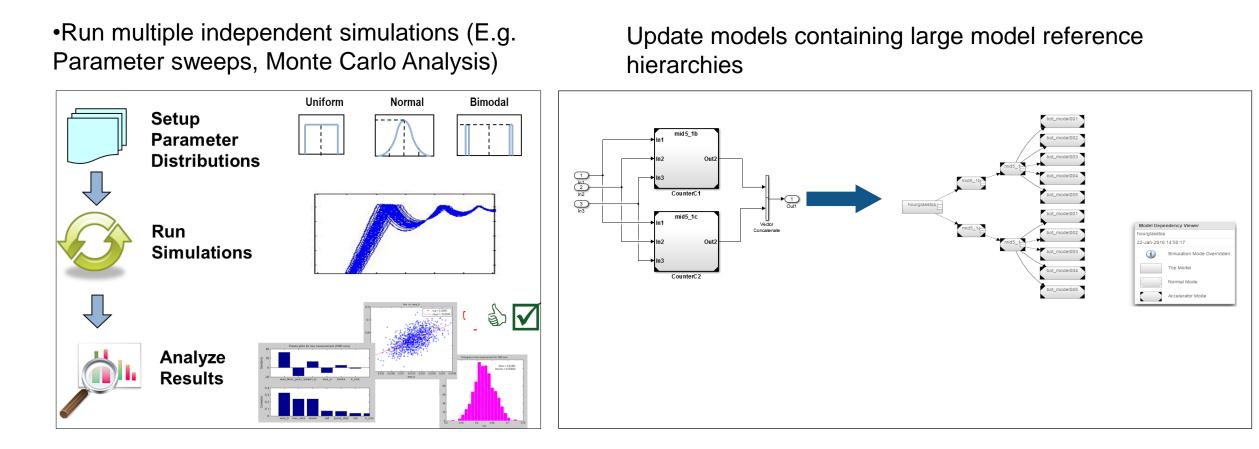
- Examples: parameter sweeps, Monte Carlo simulations
- No dependencies or communications between tasks





Leverage Parallel Computing for Simulink

Reduce the total amount of time it takes to...





Parallel Simulations using Simulink and Parsim

Directly run multiple parallel simulations from the parsim command

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We manage the parallel setup so customers can focus on their simulations

- Enables customers to easily use Simulink with parallel computing
- Simplifies customers' large simulation runs and improves their productivity
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Programming Parallel Applications – Programming Constructs

Ease of Use

Parallel-enabled toolboxes

Simple programming constructs Eg. Parfor, Batch

Advanced programming constructs spmd,createJob,labsend

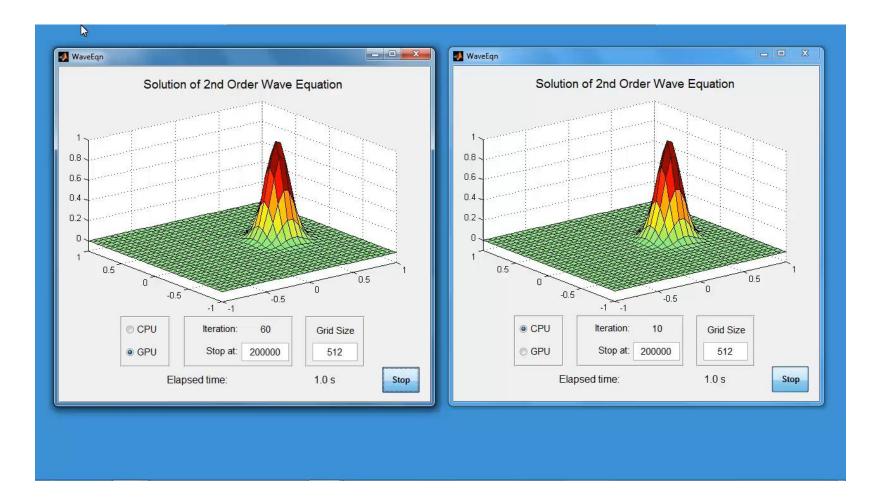


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- Accelerate applications with NVIDIA GPUs
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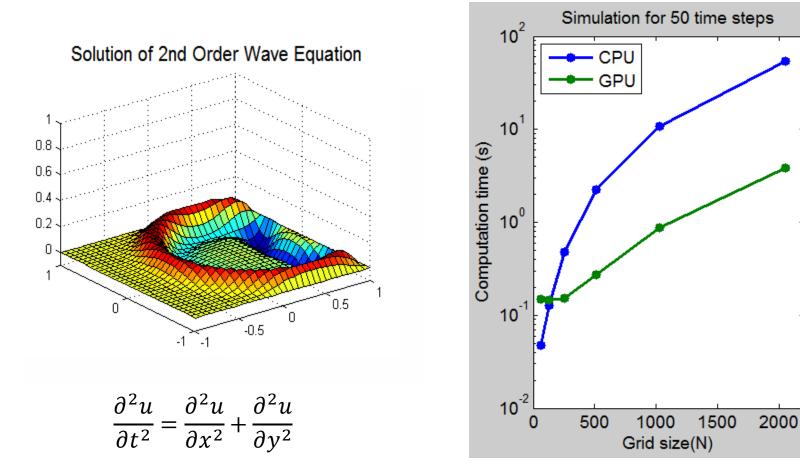


Example 3: Solving 2D Wave Equation GPU Computing





Example: Solving 2D Wave Equation GPU Computing

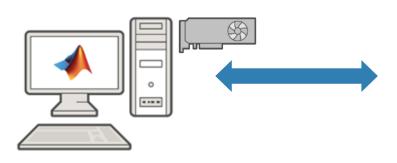


Intel Xeon Processor W3690 (3.47GHz), NVIDIA Tesla K20 GPU

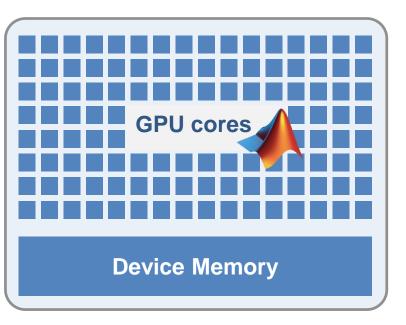


Parallel Computing Paradigm Going Parallel: GPUs

Using NVIDIA GPUs

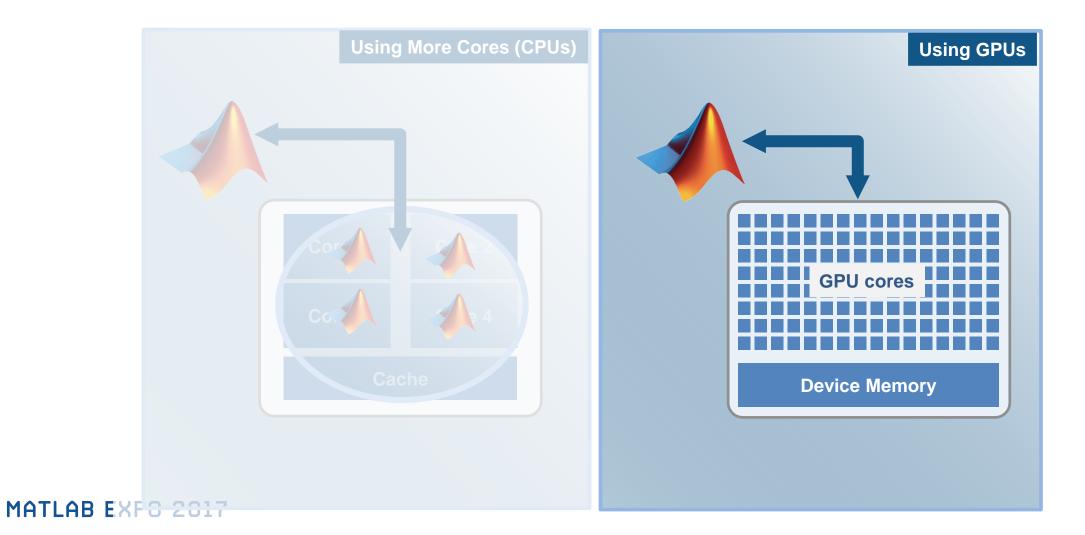


MATLAB Desktop (client)





Performance Gain with More Hardware

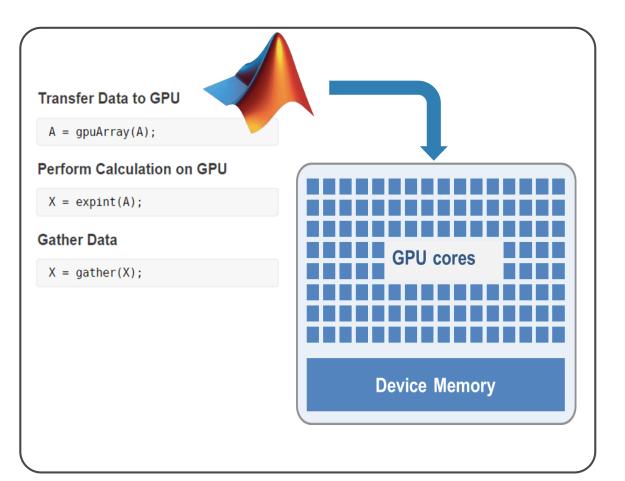




Speed-up using NVIDIA GPUs

- Ideal Problems
 - Massively Parallel and/or Vectorized operations
 - Computationally Intensive
 - Algorithm consists of supported functions
- > 300+ GPU-enabled MATLAB functions
- Additional GPU-enabled Toolboxes
 - Neural Networks
 - Image Processing
 - Communications
 - Signal Processing







^{.....} Learn More

Signal Processing – Acoustic Data Analysis NASA Langley Research

Goal: Accelerate the analysis of sound recordings from wind tunnel tests of aircraft components

Challenges

- Legacy code took 40 mins to analyze single wind tunnel test data
- Reduce processing time to make on-the-fly decisions and identify hardware problems

Why GPU Computing

Computations completed 40 times faster.

"Many operations we perform, including FFTs and matrix multiplication, are **GPU-enabled MATLAB functions**. Once we developed the initial MATLAB code for CPU execution, it took 30 minutes to get our algorithm working on the GPU **no low-level CUDA programming** was needed. The addition of GPU computing with Parallel Computing Toolbox cut it to **under a minute**, with most of that time spent on data transfer"

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Learn More

36





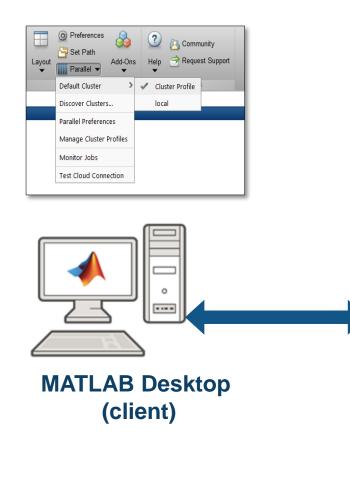


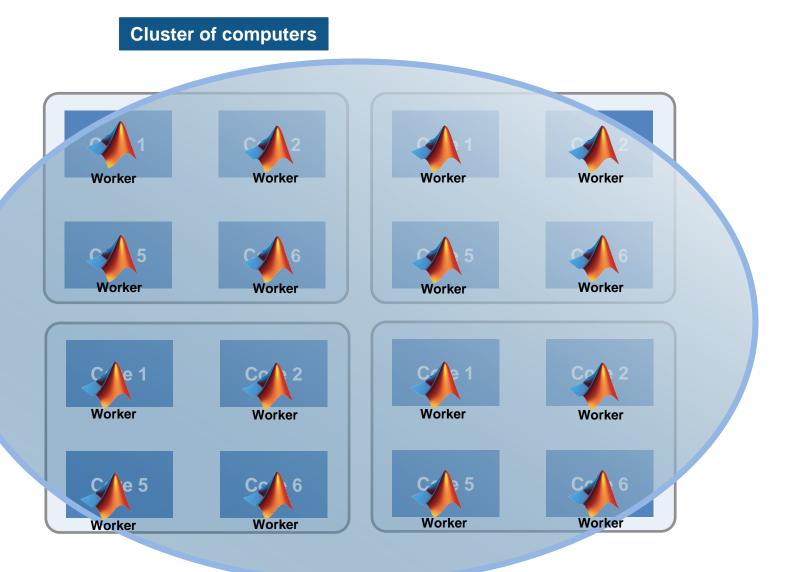
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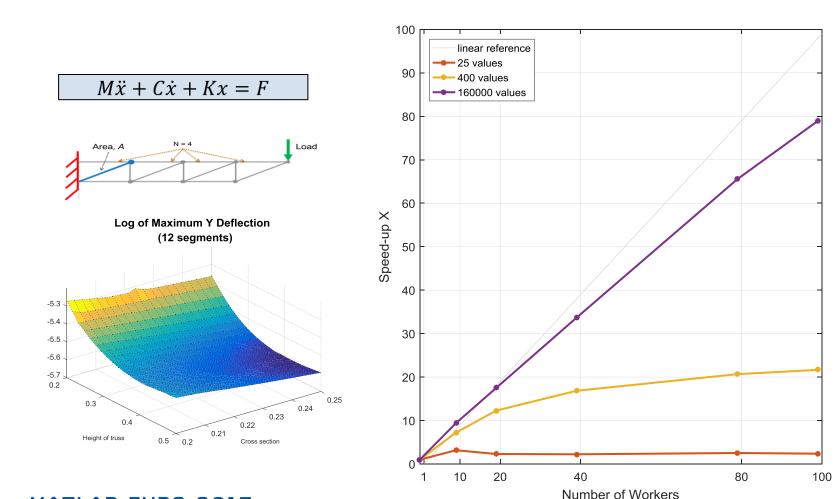
Scaling to a computer cluster







Why parallel computing matters Scaling case study with a compute cluster

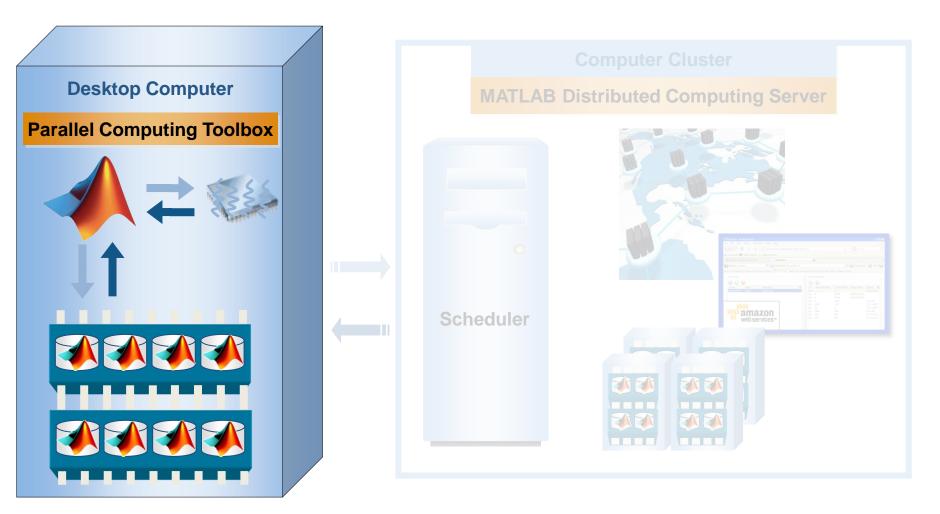


Workers	С	ompute time (minutes)	,
in pool	160e3 values	400 values	25 values
1	140	0.38	0.03
10	15	0.05	0.01
20	8.0	0.03	0.01
40	4.2	0.02	0.01
80	2.1	0.02	0.01
100	1.8	0.02	0.01

Processor: Intel Xeon E5-class v2 16 physical cores per node MATLAB R2016a

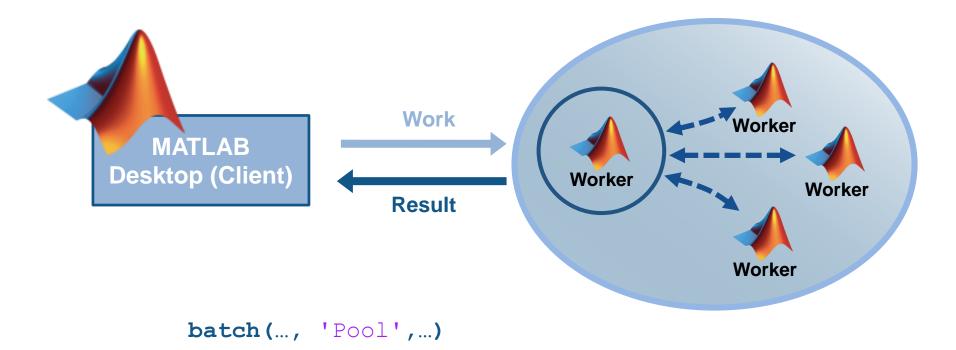


Scale Up to Clusters, Grids and Clouds





Offload and Scale Computations with batch





Scale your applications beyond the desktop











Option	Parallel Computing Toolbox	MATLAB Parallel Cloud	MATLAB Distributed Computing Server for Amazon EC2	MATLAB Distributed Computing Server for Custom Cloud	MATLAB Distributed Computing Server
Description	Explicit desktop scaling	Single-user, basic scaling to cloud	Scale to EC2 with some customization	Scale to custom cloud	Scale to clusters
Maximum workers	No limit	16	256	No limit	No limit
Hardware	Desktop	MathWorks Compute Cloud	Amazon EC2	Amazon EC2, Microsoft Azure, Others	Any
Availability	Worldwide	United States and Canada	United States, Canada and other select countries in Europe	Worldwide	Worldwide

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Learn More: Parallel Computing on the Cloud

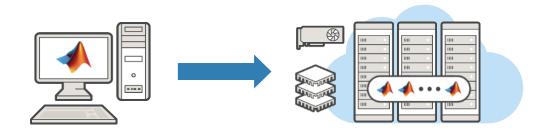


Summary and Takeaways

- Speed up your MATLAB and Simulink applications without being an expert
 - Reduce computation time by using more cores or accessing Graphical Processing Units
- Leverage the Parallel Computing Toolbox to
 - Reduce Computation Time: parfor, gpuArray, parsim
 - Offload and Scale Computations: **batch**
- Speed up Big Data Analytics
 - Using datastore and tall or distributed arrays
- Develop parallel applications on the desktop and scale to clusters seamlessly



What's new in 16b and 17a?



R2016b

- tall array support for big data
- Measure data sent to workers using ticBytes and tocBytes
- Cloud offerings with K80-equipped GPUs

R2017a

- Simplified parallel Simulink simulations using parsim
- Send data to client using DataQueue and PollableDataQueue
- Train a single deep learning network with multiple CPUs or multiple GPUs



MathWorks Training Offerings

MATLAB Programming Techniques

INTERMEDIATE

This two-day course covers details of performance optimization as well as tools for writing, debugging, and profiling code. Topics include:

- Creating robust applications
- Structuring code
- Structuring data
- Creating custom toolboxes

Prerequisites: MATLAB Fundamentals

Parallel Computing with MATLAB

INTERMEDIATE

This two-day course shows how to use Parallel Computing Toolbox[™] to speed up existing code and scale up across multiple computers using MATLAB Distributed Computing Server[™] (MDCS). Attendees who are working with long-running simulations, or large data sets, will benefit from the hands-on demonstrations and exercises in the course. Topics include:

- Parallel for-loops
- Offloading execution
- Working with clusters
- Distributing and processing large data sets
- GPU computing

Prerequisites: MATLAB Fundamentals



Learn Further

- <u>https://www.mathworks.com/solutions/parallel-computing.html</u>
- <u>https://www.mathworks.com/help/distcomp/</u>
- <u>https://www.mathworks.com/help/distcomp/examples.html</u>





Accelerating the pace of engineering and science

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Email: info@mathworks.in

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