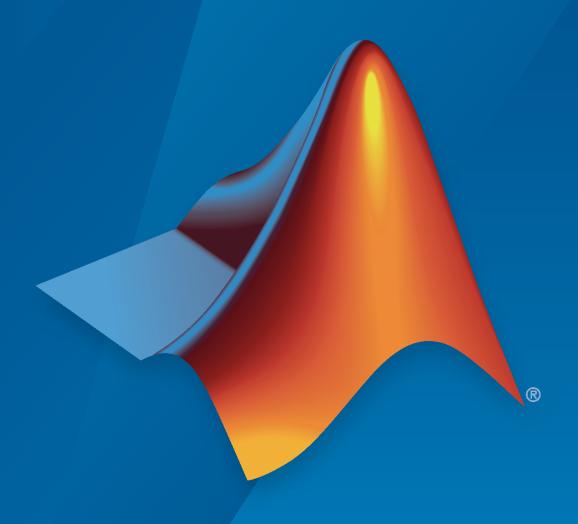
MATLAB® Parallel Server™ Release Notes



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MATLAB® Parallel Server™ Release Notes

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Contents

R2023a

PreferredPoolNumWorkers: Specify limited number of workers per cluster profile	1-2
Parallel Menu: Select a GPU using the Parallel menu	1-2
GPUDevice Object: Changes to device properties	1-3
GPU Functionality: Use functions with new and enhanced gpuArray support	1-4
PTX Build Option for mexcuda: Compile PTX files using mexcuda	1-5
Support for New GPU Architectures: Update to NVIDIA CUDA 11.8	1-5
Workflow Examples: New and updated examples and topics	1-5
Thread-Based Environment: Use new functionality on thread workers	1-6
Tall Arrays: Use functions with new and enhanced tall array support	1-6
Distributed and Tall tables and timetables: Perform calculations directly on distributed and tall tables and timetables without extracting their data	1-6
Composite arrays: Use gather to convert a Composite array into a local cell array.	1-7
Distributed Arrays: Use functions with new and enhanced distributed array support	1-8
Discover Clusters Supports Third-Party Schedulers	1-8
Third-Party Cluster Schedulers: Access simplified plugin scripts for integration on GitHub	1-8
Third-Party Cluster Schedulers: Open ports on workers to listen for connections from client	1-9
LDAP Support for MATLAB Job Scheduler Clusters: Authenticate cluster	1.9

Parallel Computing in MATLAB Online Server Environment: Speed up MATLAB Online workflows
MATLAB Parallel Server in Kubernetes: Configure and run MATLAB Parallel Server workers on Kubernetes clusters
Functionality being removed or changed
rng("default") on MATLAB parallel workers now sets random number generator settings to worker default
workers
pmode has been removed
]
Thread-Based Parallel Pool: Specify number of thread workers using parpool
parpool
parpool
parpool
parpool Parallel menu: Select Threads using the Parallel menu gpuDevice Object: Display values for byte-based memory properties w appropriate units GPU Workflow Examples: New and updated examples and topics
Parallel menu: Select Threads using the Parallel menu gpuDevice Object: Display values for byte-based memory properties w appropriate units GPU Workflow Examples: New and updated examples and topics Cloud GPU Computing: Access Cloud GPUs and Clusters in AWS Using
parpool Parallel menu: Select Threads using the Parallel menu gpuDevice Object: Display values for byte-based memory properties w appropriate units GPU Workflow Examples: New and updated examples and topics Cloud GPU Computing: Access Cloud GPUs and Clusters in AWS Using Cloud Center
parpool Parallel menu: Select Threads using the Parallel menu SpuDevice Object: Display values for byte-based memory properties wappropriate units GPU Workflow Examples: New and updated examples and topics Cloud GPU Computing: Access Cloud GPUs and Clusters in AWS Using Cloud Center New Examples: Use ValueStore to monitor batch jobs
parpool Parallel menu: Select Threads using the Parallel menu SpuDevice Object: Display values for byte-based memory properties wappropriate units GPU Workflow Examples: New and updated examples and topics Cloud GPU Computing: Access Cloud GPUs and Clusters in AWS Using Cloud Center New Examples: Use ValueStore to monitor batch jobs Data Analysis: Use mapreduce on Spark clusters Multistep Authentication for Third-Party Schedulers: Connect to remeduce on Spark clusters
parpool Parallel menu: Select Threads using the Parallel menu SpuDevice Object: Display values for byte-based memory properties wappropriate units SPU Workflow Examples: New and updated examples and topics Cloud GPU Computing: Access Cloud GPUs and Clusters in AWS Using Cloud Center New Examples: Use ValueStore to monitor batch jobs Data Analysis: Use mapreduce on Spark clusters Multistep Authentication for Third-Party Schedulers: Connect to remordients with multiple authentication options Chread-Based Environment: Use new functionality on thread workers

Distributed Arrays: Use functions with new and enhanced distributed array support	2-8
MATLAB Job Scheduler: Support for your own Java installation	2-9
Element-Wise Operations: Improved performance on GPU	2-9
svd Function: Improved performance on GPU	2-9
Functionality being removed or changed	2-10 2-10
spmdReduce, and spmdPlus	2-11 2-11
parallel.defaultProfile and parallel.listProfiles remotecopy has been removed remotemjs has been removed	2-11 2-11 2-12
R20	22a
ValueStore and FileStore objects: Retrieve data and files on MATLAB clients during job execution	3-2
Multifactor Authentication for the Generic Scheduler Interface: Connect to remote clients with RemoteClusterAccess using two or more authentication factors	3-2
Cluster resizing: Customize your MATLAB Job Scheduler cluster to resize automatically	3-2
Parallel Pools: Check status of pools	3-2
Background Pool: Number of thread workers no longer limited to 8 workers	3-2
Thread-Based Parallel Pool: See futures in active pool	3-3
cancelAll Method: Cancel currently queued and running futures in a parallel pool	3-3
GPU Functionality: Use new and enhanced gpuArray functions	3-3
Support for NVIDIA CUDA 11.2: Update to CUDA Toolkit 11.2	3-4
Tall Arrays: Use new and enhanced tall array functionality	3-4
Distributed Arrays: Use new and enhanced distributed array functionality	3-5

systems A*X = B with gpuArray for symmetric positive definite matrices A	3-5
Functionality being removed or changed	3-6 3-6 3-6
R20	21b
Parallel Language in MATLAB: Share parallel code with any MATLAB user	4-2
GPU Functionality: Use new and enhanced gpuArray functions	4-2
GPU Functionality: Use new and enhanced gpuArray functions in Statistics and Machine Learning Toolbox	4-2
GPU Functionality: Use new and enhanced gpuArray functions for working with signals and audio	4-3
Memory Usage: Use whos to check memory used by gpuArray and distributed variables	4 -4
Distributed Arrays: Use new and enhanced distributed array functionality	4-4
Thread-Based Environment: Use new and enhanced functionality on threads for working with audio, video, and images	4-4
Reference Architectures: Deploy and Run MATLAB Parallel Server from Azure Marketplace	4-6
Reference Architectures: Deploy and Run Network License Manager from Azure Marketplace	4-0
Functionality Being Removed or Changed	4-0 4-0 4-0
arrays	4- 4- 4-

GPU Functionality: Use new and enhanced gpuArray functions	5-2
GPU Functionality: Use new and enhanced gpuArray functions in Statistics and Machine Learning Toolbox	5-2
GPU Functionality: Use new and enhanced gpuArray functions for working with signals, audio, and wavelets	5-2
GPU Functionality: Use new and enhanced gpuArray functions in Image Processing Toolbox	5-3
Support for NVIDIA CUDA 11.0: Update to CUDA Toolkit 11.0	5-3
Distributed Arrays: Use new and enhanced distributed array functionality	5-4
GPU Devices: Count and compare available GPU devices	5-4
Thread-Based Parallel Pool: Use new and enhanced functionality on thread workers	5-4
parfor Examples: Use a parallel pool to speed up Monte-Carlo code	5-4
Upgrade Parallel Computing Products Together	5-4
Functionality Being Removed or Changed	5-5 5-5
{C1,,Cn} as C1,,Cn	5-5
R202	20b
GPU Functionality: Use new and enhanced gpuArray functions	6-2
GPU Functionality: Use new and enhanced gpuArray functions in Statistics and Machine Learning Toolbox	6-2
GPU Functionality: Use new and enhanced gpuArray functions for working with signals, audio, and wavelets	6-2
GPU Communications: Fast data transfer between GPUs in a parallel pool	6-3
Support for NVIDIA CUDA 10.2: Update to CUDA Toolkit 10.2	6-4
Distributed Arrays: Use new and enhanced distributed array functionality	6-4

Processing Toolbox	6-4
Array Assignment: Assign gpuArray or distributed data directly into existing MATLAB arrays	6-4
Distributed Tables: Assign data directly into existing distributed tables	6-5
Thread-Based Parallel Pool: Use a gpuArray on thread workers	6-5
parfeval Examples: Explore the state of futures and cancel them	6-5
HTCondor integration: Plugin script for HTCondor now available as an Add-On	6-5
Parallel Workflows: Compare performance of different parallel environments	6-5
Query Underlying Data: Query the underlying data type of classes	6-6
Query Parallel Functionality: Query if support for Parallel Computing Toolbox functionality is available	6-6
Improved Scalability: Use MATLAB Job Scheduler clusters with up to 4000 workers	6-6
Reference Architectures: Schedule jobs to run in AWS Batch	6-6
Docker Containers: Build a container image with a customized MATLAB installation	6-7
labSend, labReceive, labBroadcast and gop Functions: Improved performance of data transfer between GPUs in a parallel pool	6-7
Upgrade Parallel Computing Products Together	6-8
Functionality Being Removed or Changed	6-8
GPU acceleration is not available on macOS	6-8
Forward compatibility for GPU devices is disabled by default classUnderlying and isaUnderlying are not recommended	6-8 6-8
DO.)20-
K2C)20a
Parallel profiling: Learn tips and techniques to profile parallel code with new documentation	7-2
Additional Properties Documentation: Learn how to customize the behavior of the sample plugin scripts	7-2

Job Arrays: Submit job arrays to third-party schedulers with the generic scheduler interface	7-2
GPU Functionality: Use new and enhanced gpuArray functions	7-2
GPU Functionality: Use new and enhanced gpuArray functions in Statistics and Machine Learning Toolbox	7-2
GPU Functionality: New gpuArray support for spectral functions	7-3
Distributed Arrays: Use new and enhanced distributed array functionality	7-3
New Thread-Based Parallel Pool: Optimized for reduced memory usage, faster scheduling, and less data transfer, for a subset of MATLAB functions	7-3
Upgrade Parallel Computing Products Together	7-4
Functionality Being Removed or Changed	7-4 7-4 7-4
R20	19b
GPU Functionality: Use new and enhanced gpuArray functions	19b 8-2
GPU Functionality: Use new and enhanced gpuArray functions	8-2
GPU Functionality: Use new and enhanced gpuArray functions	8-2 8-2
GPU Functionality: Use new and enhanced gpuArray functions	8-2 8-2 8-2
GPU Functionality: Use new and enhanced gpuArray functions	8-2 8-2 8-2
GPU Functionality: Use new and enhanced gpuArray functions	8-2 8-2 8-2 8-2
GPU Functionality: Use new and enhanced gpuArray functions	8-2 8-2 8-2 8-2 8-2

2000 workers	8-3
Third-Party Schedulers: Check the scheduler ID of a MATLAB task	8-3
Common Job Schedulers: Improved performance of vectorized task creation	8-3
Personalized Clusters: Set Cloud Center clusters for personal use	8-4
Upgrade Parallel Computing Products Together	8-
Functionality Being Removed or Changed	8- 8- 8-
R20	19
Updated license model: Scale beyond 200 workers without checking out more licenses	9-
Renamed Product: MATLAB Distributed Computing Server renamed to MATLAB Parallel Server	9-
parfor-Loops: Use parfor without a parallel pool	9-
Automatic Cluster Resizing: Resize Cloud Center clusters on Amazon based on usage	9-
Benchmarks: Use new examples to evaluate the performance of your cluster	9-
parpool Resiliency: Parallel pools are now resilient to dropped network connections	9-
Parallel Deep Learning Workflows: Explore deep learning with multiple-GPUs	9-
Custom Datastore: Read from Hadoop based databases using the custom datastore framework	9-
GPU Functionality: Use new and enhanced gpuArray functions	9-
Support for NVIDIA CUDA 10.0: Update to CUDA Toolkit 10.0	9-
Distributed Arrays: Use new and enhanced distributed array functionality	9-
Reference Architectures: Support for network license manager	9-

Support for MPICH: Update to MPICH Version 3 on Linux for third-party schedulers	9-4
Upgrade Parallel Computing Products Together	9-4
Functionality Being Removed or Changed	9-4 9-4
future release	9-5
release	9-5 9-5 9-5 9-6
R20)18b
GPU Support: View details of GPU support on over 600 function pages, and browse GPU support for functions by toolbox	10-2
GPU Functionality: Use new and enhanced gpuArray functions, such as spline interpolation and vecnorm	10-2
Support for NVIDIA CUDA 9.1: Update to CUDA Toolkit 9.1	10-2
GPU Optimizations: Enhanced support and optimizations for GPU	10-2
GPU Examples: Explore GPU workflows on single and multiple GPUs	10-2
Distributed Arrays: Use new and enhanced distributed array functionality including support for vecnorm and writing to Amazon S3 and Azure	, 10-3
	10-3
Distributed Support: View details of distributed array support on over 400 function pages, and browse distributed support for functions by toolbox	10-3
Parallel Extended Capabilities: View details of automatic parallel support on function pages, and browse support for functions by toolbox	10-3
Write Cloud Data: Write distributed arrays and tall arrays to Amazon S3 and Windows Azure storage	10-4
Big Data in the Cloud: Explore MATLAB capabilities for big data	10-4
Streamlined Cloud Cluster Setup: Create new Cloud Center clusters directly from the MATLAB desktop	10-4
Improved Scalability: Use up to 1024 workers per parallel pool	10-4

Reference Architectures: Create customized MATLAB Distributed Computing Server clusters in the cloud using Amazon Web Services (AWS) or Microsoft Azure	10-4
Cluster Workflows: Learn how to scale up from desktop to cluster	10-4
Streamlined Installation Documentation: Simplified workflows for integrating MATLAB Distributed Computing Server in your cluster	10-5
Generic Scheduler Documentation: Set up Schedulers Using Improved Instructions	10-5
Parallel Deep Learning Workflows: Explore deep learning with multiple GPUs locally or in the cloud	10-5
Upgrade Parallel Computing Products Together	10-5
R20	018a
Parfeval callbacks: New afterAll and afterEach methods for parallel	
futures	11-2
Slurm support: Slurm is a fully supported scheduler	11-2
Support for NVIDIA Volta: Update to CUDA 9, support for Volta class GPUs	11-2
Improved file mirroring: Performance of file mirroring for generic scheduler integration	11-2
Parfor performance improvements: More efficient broadcast variables in parfor for non-local clusters	11-2
GPU Array Support: Use enhanced gpuArray functions	11-2
Distributed Array Support: Use enhanced distributed array functions	11-2
Parameter Sweep Example: Use a DataQueue to monitor results during computations on a parallel pool	11-3
Discontinued Support for GPU Devices of Compute Capability less than 3.0	11-3
Backwards Compatibility for MATLAB Job Scheduler: Upgrade MATLAB Job Scheduler clusters, and continue to use previous releases of Parallel Computing Toolbox	11-3
Upgrade Parallel Computing Products Together	11-3

Functionality Being Removed or Changed	11-4
i directionality Being Itemoved of Changed	

R2017b

Improved Parallel Language Performance: Execute parallel language constructs with reduced overhead	12-2
Tall Array Support: Use tall arrays with Windows client access to Linux Spark clusters	12-2
Improved Parallel Pool Robustness: Run pools without Message Passing Interface (MPI) by default, making pools resilient to workers crashing	12-2
Improved MATLAB Integration with Third-Party Schedulers: Use the Generic Profile Wizard for easier installation and setup of MATLAB Distributed Computing Server	12-2
Cloud Storage: Work with data in Windows Azure Blob Storage	12-2
Copy Client Environment to Workers on Any Cluster: Specify which environment variables on your client machine your workers should automatically inherit	12-2
Client Path Sharing: Add user-added-entries on the client's path to the workers' paths	12-3
GPU Array Support: Use enhanced gpuArray functions	12-3
Distributed Array Support: Use enhanced distributed array functions	12-3
Enhanced Support for Microsoft Windows HPC Pack	12-3
Discontinued Support for GPU Devices of Compute Capability less than 3.0	12-3
Backwards Compatibility for MATLAB Job Scheduler: Upgrade MATLAB Job Scheduler clusters, and continue to use previous releases of Parallel Computing Toolbox	12-4
Upgrade Parallel Computing Products Together	12-4
Functionality Being Removed or Changed	12-4

Backwards Compatibility for MATLAB Job Scheduler: Upgrade MATLAB Job Scheduler clusters, and continue to use previous releases of Parallel Computing Toolbox	13-2
Simplified Integration for Third-Party Cluster Schedulers: Updates to generic scheduler integration allow folder-based configuration and eliminate the need to specify function handles	13-2
Ability to Re-create Parallel Jobs: Easily rerun all failed or cancelled task for a job	s 13-2
More Responsive Job Monitor: Automatic updates for new, submitted, or deleted jobs or tasks	13-2
Access to Intermediate Results and Updates in Parallel Computations: Poll for messages or data from different workers during parallel workflows	13-2
	10 =
Distributed Array Support: Use enhanced distributed array functions	13-2
Support for Spark 2.x enabled Hadoop clusters	13-3
Discontinued Support for GPU Devices of Compute Capability less than 3.0	13-3
Upgrade parallel computing products together	13-3
Properties of generic cluster objects have changed	13-3
Functionality Being Removed or Changed	13-4
R2	016b
Backwards Compatibility: Upgrade MATLAB Job Scheduler clusters, and	
continue to use the previous release of Parallel Computing Toolbox	14-2
Cluster Profile Validation: Choose which validation stages run and the number of MATLAB workers to use	14-2
Parallel Support for Tall Arrays: Process big data with tall arrays in parallel on your desktop, MATLAB Distributed Computing Server, and Spark clusters	14-2
Parallel Menu Enhancement: Use the new menu items in the Parallel Menu to configure and manage cloud based resources	14-2

New Data Types in Distributed Arrays: Use enhanced functions for creating distributed arrays of: datetime; duration; calendarDuration; string; categorical; and table	14-2
Loading Distributed Arrays: Load distributed arrays in parallel using datastore	14-4
datetime Support for Timestamps: Use built-in datetime objects in MATLAB to access timestamp information for jobs and tasks	14-4
Data Transfer Measurement: Use ticBytes and tocBytes to measure the data transfer between MATLAB workers in a parallel pool	14-4
Multithreaded Workers: Use multiple computational threads on your MATLAB workers	14-4
MathWorks Hosted License Manager: Use new option in MATLAB Distributed Computing Server script to ensure workers use MathWorks Hosted License Manager	14-5
JSON support for nodestatus: View the output of the nodestatus script in JavaScript Object Notation (JSON) format	14-5
Upgrade Parallel Computing Products Together	14-5
Functionality Being Removed or Changed	14-5
R20	16a
Support for Distributed Arrays: Use enhanced distributed array functions including sparse input to direct (mldivide) and iterative solvers (cgs and pcg)	15-2
Hadoop Kerberos Support: Improved support for Hadoop in a Kerberos authenticated environment	15-2
Transfer unlimited data between client and workers, and attached files up to 4GB in total, in any job using a MATLAB Job Scheduler cluster	15-2
Third Party Scheduler Integration: Obtain integration scripts for Third Party Schedulers (IBM Platform LSF, Grid Engine, PBS and SLURM) from MATLAB Central File Exchange instead of Parallel Computing	
Toolbox	15-2

Discontinued support for parallel computing products on 32-bit Windows	
operating systems	16-2
Scheduler integration scripts for SLURM	16-2
Improved performance of mapreduce on Hadoop 2 clusters	16-2
parallel.pool.Constant function to create constant data on parallel pool workers, accessible within parallel language constructs such as parfor and parfeval	16-2
Upgrade parallel computing products together	16-2
DO.	04 F -
K20	015a
Support for mapreduce function on any cluster that supports parallel pools	17-2
Using DNS for cluster discovery	17-2
MS-MPI support for MATLAB Job Scheduler clusters	17-2
Ports and sockets in mdce_def file	17-2
Discontinued support for GPU devices on 32-bit Windows computers	17-2
Discontinued support for parallel computing products on 32-bit Windows computers	17-3
R20)14b
Data Analysis on Hadoop clusters using mapreduce	18-2
Additional MATLAB functions for distributed arrays, including fft2, fftn, ifft2, ifftn, cummax, cummin, and diff	18-2

Duplication of an existing job, containing some or all of its tasks	19-2
More MATLAB functions enhanced for distributed arrays	19-2
Old Programming Interface Removed	19-2
matlabpool Function Being Removed	19-3
R20	013b
parpool: New command-line interface (replaces matlabpool), desktop indicator, and preferences for easier interaction with a parallel pool of MATLAB workers	20-2
Automatic start of a parallel pool when executing code that uses parfor o spmd	or 20-2
Option to start a parallel pool without using MPI	20-2
More MATLAB functions enabled for distributed arrays: permute, ipermute, and sortrows	20-3
Upgraded MPICH2 Version	20-3
Discontinued Support for parallel.cluster.Mpiexec	20-3
R2	013a
Automatic detection and transfer of files required for execution in both batch and interactive workflows	21-2
R20	012b
Automatic detection and selection of specific GPUs on a cluster node when multiple GPUs are available on the node	22-2
Detection of MATLAB Distributed Computing Server clusters that are available for connection from user desktops through Profile Manager	22-2

R2023a

Version: 7.8

New Features

Compatibility Considerations

PreferredPoolNumWorkers: Specify limited number of workers per cluster profile

The global preference for the default number of workers in a pool is replaced by a per-profile property. You can configure the new PreferredPoolNumWorkers cluster object property in the cluster profile manager or at the command line. The PreferredPoolNumWorkers property default is the number of workers available to the cluster (NumWorkers) for personal cloud cluster profiles or the minimum of NumWorkers and 32 for other nonlocal profiles.

Use this new property to create pools with a reasonable number of workers when you call parpool without requesting a specific number of workers. You can also use this property to control how many workers you start on clusters to avoid taking up too many resources.

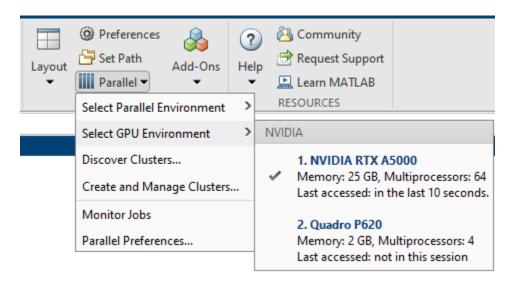
For local profiles, the default parallel pool size is NumWorkers. For all profiles, you can continue to request a specific number of workers when you call parpool to start a parallel pool outside the main body of code. If you specify a pool size at the command line, you override your preferences. For more details, see "Pool Size and Cluster Selection" (Parallel Computing Toolbox).

Compatibility Considerations

The **Preferred number of workers in a parallel pool** option in **Parallel Preferences** has been removed. Use the PreferredPoolNumWorkers or NumWorkers cluster object properties instead. For more information, see Functionality being removed or changed on page 1-10.

Parallel Menu: Select a GPU using the Parallel menu

You can now select which GPU device to use for computation from the MATLAB desktop. On the **Home** tab, in the **Environment** area, select **Parallel** > **Select GPU Environment**. For each available GPU, the menu displays the total memory, the multiprocessor count, and an indication of when the device was last used. To inspect more properties of your GPU devices, use the <code>gpuDevice</code> function.



GPUDevice Object: Changes to device properties

gpuDevice objects that the gpuDevice function returns now have these additional properties:

- GraphicsDriverVersion Graphics driver version currently in use.
- DriverModel Operating model of the graphics driver.

On Windows® operating systems, the model options are:

- 'WDDM' Display model.
- 'TCC' Compute model.

On other operating systems, DriverModel is 'N/A'.

• CachePolicy — Policy that determines how much GPU memory the GPU can cache to accelerate computation. You can set the CachePolicy property to "balanced", "minimum", or "maximum".

For example, select a GPU using the <code>gpuDevice</code> function to inspect the new properties.

```
D = gpuDevice
D =
  CUDADevice with properties:
                       Name: 'NVIDIA RTX A5000'
                      Index: 1
         ComputeCapability: '8.6'
            SupportsDouble: 1
     GraphicsDriverVersion: '511.79'
DriverModel: 'TCC'
            ToolkitVersion: 11.2000
        MaxThreadsPerBlock: 1024
          MaxShmemPerBlock: 49152 (49.15 KB)
        MaxThreadBlockSize: [1024 1024 64]
               MaxGridSize: [2.1475e+09 65535 65535]
                  SIMDWidth: 32
               TotalMemory: 25553076224 (25.55 GB)
           AvailableMemory: 25145376768 (25.15 GB)
               CachePolicy: 'balanced'
       MultiprocessorCount: 64
              ClockRateKHz: 1695000
               ComputeMode: 'Default'
      GPUOverlapsTransfers: 1
    KernelExecutionTimeout: 0
          CanMapHostMemory: 1
           DeviceSupported: 1
           DeviceAvailable: 1
            DeviceSelected: 1
```

Change the caching policy to allow the GPU to cache the maximum amount of memory for accelerating computation.

```
D.CachePolicy = "maximum";
```

The DriverVersion property no longer displays by default, but you can still query the property using dot notation. There are no plans to remove the DriverVersion property.

GPU Functionality: Use functions with new and enhanced gpuArray support

These MATLAB functions have new and enhanced gpuArray support:

- cumtrapz
- eig You can now provide two input matrices, compute three outputs (eigenvalues, left
 eigenvectors, and right eigenvectors), specify the output format of the eigenvalues as a matrix or
 a vector, and provide input matrices containing NaN or Inf values.
- fillmissing
- interp1 You can now use the "next" and "previous" interpolation methods.
- 1scov
- mustBeFloat, mustBeGreaterThan, mustBeGreaterThanOrEqual, mustBeInRange, mustBeLessThan, mustBeLessThanOrEqual, mustBeNegative, mustBeNonempty, mustBeNonmissing, mustBeNonnegative, mustBeNonpositive, mustBeNonsparse, mustBeNumeric, mustBeNumericOrLogical, mustBePositive, mustBeReal, mustBeScalarOrEmpty, mustBeVector
- tensorprod

For more information, see "Run MATLAB Functions on a GPU" (Parallel Computing Toolbox).

These Statistics and Machine Learning $Toolbox^{TM}$ functions have new and enhanced gpuArray support:

- regress
- fitrsvm You can now specify gpuArray inputs.
 - Most object functions of RegressionSVM, CompactRegressionSVM, and RegressionPartitionedSVM run on a GPU when you provide GPU array input arguments. These object functions do not support GPU array input arguments.

Object Function	Model Type
incrementalLearner	RegressionSVM and CompactRegressionSVM
lime	RegressionSVM and CompactRegressionSVM
shapley	RegressionSVM and CompactRegressionSVM
update	CompactRegressionSVM

 You can now use the gather function to create a RegressionSVM, CompactRegressionSVM, or RegressionPartitionedSVM object from an equivalent object fitted with GPU array data. The properties of the new object are stored in host memory.

For a list of all Statistics and Machine Learning Toolbox functions with GPU functionality, see Functions with ${\tt gpuArray}$ support.

These Signal Processing Toolbox™ functions have new and enhanced gpuArray support:

- bandpower
- medfreq
- obw
- powerbw
- rssq
- signalFrequencyFeatureExtractor
- signalTimeFeatureExtractor
- sinad
- snr
- thd

For a list of all Signal Processing Toolbox functions with GPU functionality, see Functions with gpuArray support.

These Wavelet Toolbox™ functions have new and enhanced gpuArray support:

- modwpt
- wentropy

For a list of all Wavelet Toolbox functions with GPU functionality, see Functions with gpuArray support.

PTX Build Option for mexcuda: Compile PTX files using mexcuda

You can now compile parallel thread execution (PTX) files using the mexcuda function by specifying the -ptx build option. The function compiles the PTX file and gives it the same name as the input $CUDA^{\otimes}$ C++ (CU) source file.

You no longer need to install the CUDA Toolkit to compile a PTX file from a CU file.

For an updated overview of the workflow for creating and executing your own custom CUDAKernel object, see "Run CUDA or PTX Code on GPU" (Parallel Computing Toolbox).

Support for New GPU Architectures: Update to NVIDIA CUDA 11.8

Parallel Computing Toolbox™ software now uses CUDA version 11.8, which supports NVIDIA® GPUs with compute capability up to 9.x, including Hopper architecture GPUs. For more information, see "GPU Computing Requirements" (Parallel Computing Toolbox).

Workflow Examples: New and updated examples and topics

These new and updated examples show how to accelerate your code using a GPU:

- "Accelerate Linear Model Fitting on GPU" (Statistics and Machine Learning Toolbox)
- "Improve Performance of Small Matrix Problems on the GPU Using pagefun" (Parallel Computing Toolbox)

This updated example shows how to monitor the training of deep neural networks in a batch job using a ValueStore object:

• "Send Deep Learning Batch Job to Cluster" (Deep Learning Toolbox)

This new topic helps you choose the right parallel language and workflow for your needs:

• "Parallel Language Decision Tables" (Parallel Computing Toolbox)

Thread-Based Environment: Use new functionality on thread workers

These MATLAB functions and classes now have thread-based support:

- readtable
- ToFile
- ToUniqueFile

For more information, see Run MATLAB Functions in Thread-Based Environment.

Tall Arrays: Use functions with new and enhanced tall array support

These functions have new and enhanced tall array support:

- mape
- rmse

For more information, see "Tall Arrays" (MATLAB).

Distributed and Tall tables and timetables: Perform calculations directly on distributed and tall tables and timetables without extracting their data

You can now perform calculations directly on distributed and tall tables and timetables without extracting their data. In previous releases, all calculations require you to extract data from your distributed and tall tables and timetables by indexing into them.

For example, to scale a distributed table that contains numeric data, multiply it by a scale factor.

V1	V2
6.9483	9.5022 0.34446

These MATLAB functions have new support for performing calculations directly on distributed and tall tables and timetables.

- Arithmetic functions ceil, fix, floor, ldivide, minus, mod, plus, power, prod, rdivide, rem, round, sum, times
- Relational functions eq, ge, gt, le, lt, ne
- Boolean functions and, not, or, xor
- Trigonometric functions acos, acosd, acosh, acot, acotd, acoth, acsc, acscd, acsch, asec, asecd, asin, asind, asinh, atan2, atan2d, atand, atanh, cos, cosd, cosh, cospi, cot, cotd, cscd, csch, sec, secd, sech, sin, sind, sinh, sinpi, tan, tand, tanh
- Exponential and logarithmic functions exp, expm1, log, log10, log1p, log2, nextpow2, nthroot, pow2, reallog, realpow, realsqrt, sqrt
- Complex number functions abs
- Statistics functions max, mean, min

For more information about performing calculations directly on tables and timetables, see "Direct Calculations on Tables and Timetables" (MATLAB).

Composite arrays: Use gather to convert a Composite array into a local cell array.

You can now use the gather function to convert a Composite array containing data on parallel workers into a cell array in the local workspace.

For example, gather data after the execution of an spmd statement.

```
parpool("Threads",3)
spmd, c = ones(spmdIndex); end
data = gather(c)

data =
   1×3 cell array
   {[1]}   {2×2 double}   {3×3 double}
```

Compatibility Considerations

In previous releases, when you use gather on a Composite array, MATLAB returns the same Composite array as the output. If you have code that relies on this behaviour, update your code to avoid compatibility issues.

Distributed Arrays: Use functions with new and enhanced distributed array support

These functions have new and enhanced distributed array support:

- cumtrapz
- fillmissing
- kron
- lscov
- movmad, movmax, movmean, movmedian, movmin, movprod, movstd, movsum, movvar You can now use the SamplePoints name-value argument.
- mpower You can now use a square matrix for the base or exponent.
- mustBeFinite, mustBeFloat, mustBeGreaterThan, mustBeGreaterThanOrEqual, mustBeInRange, mustBeInteger, mustBeLessThan, mustBeLessThanOrEqual, mustBeNegative, mustBeNonempty, mustBeNonmissing, mustBeNonNan, mustBeNonnegative, mustBeNonpositive, mustBeNonsparse, mustBeNonzero, mustBeNumeric, mustBeNumericOrLogical, mustBePositive, mustBeReal, mustBeScalarOrEmpty, mustBeVector
- null
- tensorprod

For more information, see "Run MATLAB Functions with Distributed Arrays" (Parallel Computing Toolbox).

Discover Clusters Supports Third-Party Schedulers

The Discover Clusters dialog can now locate third-party scheduler clusters for different releases of parallel computing products. For information about cluster discovery, see "Discover Clusters" (Parallel Computing Toolbox).

Third-Party Cluster Schedulers: Access simplified plugin scripts for integration on GitHub

When you integrate MATLAB with your scheduler using MATLAB Parallel Server[™], you can now use a single set of scripts and specify the values of one of these AdditionalProperties properties to describe your network configuration.

Network Configuration	Previous Submission Mode	New Properties to Set
The MATLAB client does not share the file system with the cluster nodes.	Shared	• HasSharedFilesystem = true
The MATLAB client is unable to directly submit jobs to the third-		• HasSharedFilesystem = true
party scheduler.		• ClusterHost = HeadnodeHostname

Network Configuration	Previous Submission Mode	New Properties to Set
The MATLAB client does not share the file system with the cluster nodes and is unable to directly submit jobs to the third-party scheduler.	Nonshared	 HasSharedFilesystem = false ClusterHost = HeadnodeHostname

In previous releases, you have to choose a set of scripts for shared, nonshared, and remote submission modes

The simplified plugin scripts for these third-party schedulers are now available in these GitHub® repositories:

- SLURM
- LSF
- · Grid Engine family
- · PBS family
- HTCondor

You can still download the plugin scripts from MATLAB Central™ File Exchange.

Third-Party Cluster Schedulers: Open ports on workers to listen for connections from client

You can now open listening ports on MATLAB Parallel Server workers running on third-party scheduler clusters to enable your MATLAB client to connect to the workers. Use this functionality to run interactive parallel pools on third-party scheduler clusters without the need to modify your network systems such as firewalls. For more details, see pctconfig.

LDAP Support for MATLAB Job Scheduler Clusters: Authenticate cluster logins against LDAP credentials

Validate and control user access to a MATLAB Job Scheduler cluster by using a Lightweight Directory Access Protocol (LDAP) server to authenticate user logins. For more information, see "Configure LDAP Server Authentication for MATLAB Job Scheduler".

Cluster Security: Restrict use of MATLAB Job Scheduler cluster commands

You can now restrict the use of cluster changing commands to specific users by verifying commands sent to MATLAB Job Scheduler clusters. In previous releases, any user can execute commands that can change the state of cluster. For more information, see "Set Cluster Command Verification".

Parallel Computing in MATLAB Online Server Environment: Speed up MATLAB Online workflows

You can now integrate Parallel Computing Toolbox software and MATLAB Parallel Server with your MATLAB Online Server $^{\text{\tiny TM}}$ instance and use parallel computing to speed up MATLAB Online $^{\text{\tiny TM}}$

workflows. For more details about how to configure and use parallel computing features in your MATLAB Online Server instance, see Integrate MATLAB Online Server with Parallel Computing Toolbox and MATLAB Parallel Server and Use Parallel Computing in MATLAB Online Hosted by Your Organization.

MATLAB Parallel Server in Kubernetes: Configure and run MATLAB Parallel Server workers on Kubernetes clusters

You can now integrate MATLAB Parallel Server with your Kubernetes® cluster and interface Parallel Computing Toolbox software on your computer to the Kubernetes cluster. For more details, see the Parallel Computing Toolbox for MATLAB Parallel Server with Kubernetes plugin script on GitHub or MATLAB Central File Exchange.

Functionality being removed or changed

Preferred number of workers in a parallel pool option has been removed Behavior change

The **Preferred number of workers in a parallel pool** option in **Parallel Preferences** has been removed. This change gives you more control over how many workers to start on nonlocal clusters.

For MATLAB Job Scheduler, third party schedulers, and cloud clusters, you can use the PreferredPoolNumWorkers property of your cluster profiles to specify the preferred number of workers to start in a parallel pool. For the local Processes and Threads profiles, use the NumWorkers property to specify the default number of workers to start in a parallel pool.

rng("default") on MATLAB parallel workers now sets random number generator settings to worker default

Behavior change

When you call the rng function with the "default" argument on MATLAB parallel workers, MATLAB resets the random number generator settings to the worker default values. The default corresponds to the Threefry generator with 20 rounds and a seed of 0

In previous releases, when you run rng("default") on parallel workers, MATLAB changes the worker random number generator settings to the client default values. The default corresponds to the Mersenne Twister generator with a seed of 0.

parallel.pool.Constant objects no longer automatically transferred to workers Behavior change

MATLAB no longer automatically transfers the parallel.pool.Constant object from your current MATLAB session to workers in a parallel pool. MATLAB sends the Constant object to workers only if the object is required to execute your code.

Thread-Based Parallel Pool: startup or matlabrc no longer runs when workers start Behavior change

When you start a thread-based parallel pool, the thread-based workers no longer run the startup or matlabrc file. All startup options on the client are automatically mirrored to the thread-based workers.

pmode has been removed

Errors

The pmode function has been removed. To execute commands interactively on multiple workers, use spmd instead.

pload and psave have been removed

Errors

The pload and psave functions have been removed. To save and load data on workers, in the form of composite arrays or distributed arrays, use dsave and dload instead.

R2022b

Version: 7.7

New Features

Compatibility Considerations

Thread-Based Parallel Pool: Specify number of thread workers using parpool

You can now specify the pool size of a thread-based parallel pool by using parpool (Parallel Computing Toolbox).

For example, start a pool of four thread workers.

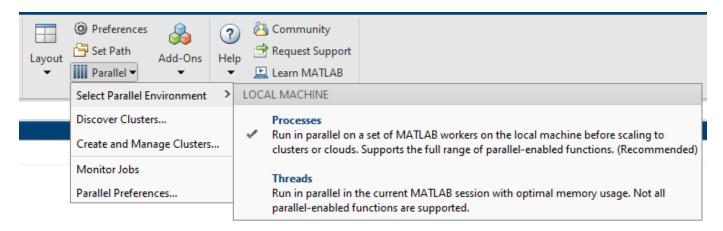
```
parpool("Threads",4);
```

You can also set Threads as the default parallel environment and start a parallel pool with the largest possible number of workers.

```
parallel.defaultProfile("Threads");
parpool([1 50]);
```

Parallel menu: Select Threads using the Parallel menu

You can now select the Threads parallel environment option from the MATLAB desktop **Parallel** menu.



- You can set the default parallel environment on your local machine from the MATLAB desktop **Home** tab, in the **Environment** area, by selecting one of these options:
 - Parallel > Select Parallel Environment > Processes
 - Parallel > Select Parallel Environment > Threads

You can also change the default profile in **Parallel Preferences**. For more information, see Specify Your Parallel Preferences (Parallel Computing Toolbox).

Compatibility Considerations

The local profile is no longer recommended. For more information, see Functionality being removed or changed (Parallel Computing Toolbox).

gpuDevice Object: Display values for byte-based memory properties with appropriate units

The gpuDevice (Parallel Computing Toolbox) object now displays the TotalMemory, AvailableMemory, and MaxShmemPerBlock properties with appropriately scaled units. The object displays these properties as bytes (B), kilobytes (KB), megabytes (MB), or gigabytes (GB). Accessing any of these properties using dot notation still returns a value in bytes.

For example, create a gpuDevice (Parallel Computing Toolbox) object and access its TotalMemory property.

```
g = gpuDevice
 CUDADevice with properties:
                      Name: 'NVIDIA RTX A5000'
                     Index: 1
         ComputeCapability: '8.6'
            SupportsDouble: 1
             DriverVersion: 11.6000
            ToolkitVersion: 11.2000
        MaxThreadsPerBlock: 1024
          MaxShmemPerBlock: 49152 (49.15 KB)
        MaxThreadBlockSize: [1024 1024 64]
               MaxGridSize: [2.1475e+09 65535 65535]
                 SIMDWidth: 32
               TotalMemory: 25553076224 (25.55 GB)
           AvailableMemory: 25153765376 (25.15 GB)
      MultiprocessorCount: 64
              ClockRateKHz: 1695000
               ComputeMode: 'Default'
      GPUOverlapsTransfers: 1
   KernelExecutionTimeout: 0
          CanMapHostMemory: 1
           DeviceSupported: 1
           DeviceAvailable: 1
           DeviceSelected: 1
g.TotalMemory
ans =
  2.5553e+10
```

GPU Workflow Examples: New and updated examples and topics

These new and updated examples and topics show how to accelerate your code using a GPU:

- Run MATLAB using GPUs in the Cloud (Parallel Computing Toolbox)
- Measure and Improve GPU Performance (Parallel Computing Toolbox)
- Improve Performance Using a GPU and Vectorized Calculations (Parallel Computing Toolbox)
- Improve Performance of Element-Wise MATLAB Functions on the GPU Using arrayfun (Parallel Computing Toolbox)

• Using GPU arrayfun for Monte-Carlo Simulations (Parallel Computing Toolbox)

Cloud GPU Computing: Access Cloud GPUs and Clusters in AWS Using Cloud Center

If you do not have a GPU available, you can speed up your MATLAB code with one or more high-performance NVIDIA GPUs in the cloud. Working in the cloud requires some initial setup, but using cloud resources can significantly accelerate your code without requiring you to buy and set up your own local GPUs. The simplest way to access a high-performance cloud GPU is to use Cloud Center.

From April 2022, you can use Cloud Center to start MATLAB in Amazon® Web Services (AWS®). You can start a single machine, with MATLAB installed, that you can access from a web browser or a remote desktop application. To get started, see Get Started with Cloud Center and Start MATLAB on Amazon Web Services (AWS) Using Cloud Center. To learn more about other cloud options for GPU access, see Run MATLAB using GPUs in the Cloud (Parallel Computing Toolbox).

You can continue to use Cloud Center to start and manage MATLAB Parallel Server clusters that you can access from any MATLAB client. MATLAB Parallel Server clusters provide more hardware, including GPU and CPU clusters.

New Examples: Use ValueStore to monitor batch jobs

These new examples show how to retrieve data and monitor batch jobs with a ValueStore object:

- Monitor Batch Jobs with ValueStore (Parallel Computing Toolbox) shows how to use ValueStore to update a waitbar.
- Monitor Monte Carlo Batch Jobs with ValueStore (Parallel Computing Toolbox) shows how to use ValueStore to access and monitor data from a batch Monte Carlo simulation.

Data Analysis: Use mapreduce on Spark clusters

Parallel Computing Toolbox and MATLAB Parallel Server support the use of Spark $^{\text{\tiny TM}}$ clusters for the execution environment of mapreduce applications. For more information, see:

- · Configure a Spark Cluster
- Use Tall Arrays on a Spark Cluster (Parallel Computing Toolbox)

Multistep Authentication for Third-Party Schedulers: Connect to remote clients with multiple authentication options

You can now use a RemoteClusterAccess (Parallel Computing Toolbox) object to perform multistep authentication with any number of authentication options. To connect to a cluster with multiple authentication requirements, specify the AuthenticationMode property of the RemoteClusterAccess object as a string array or cell array containing a combination of "Agent", "IdentityFile", "Multifactor", and "Password".

For example, to connect to a cluster that requires a password and identity file, you can specify a password and identity file name-value argument as a cell array:

RemoteClusterAccess(AuthenticationMode={"IdentityFile","Password"});

The sample plugin scripts for third-party schedulers now accept the cell array {"IdentityFile", "Password"} as the value of the AuthenticationMode property of an AdditionalProperties object, which sets the same AuthenticationMode property value in the RemoteClusterAccess object.

Thread-Based Environment: Use new functionality on thread workers

These Parallel Computing Toolbox functions now have new thread-based support:

- spmdBarrier (Parallel Computing Toolbox)
- spmdBroadcast (Parallel Computing Toolbox)
- spmdIndex (Parallel Computing Toolbox)
- spmdProbe (Parallel Computing Toolbox)
- spmdReceive (Parallel Computing Toolbox)
- spmdSend (Parallel Computing Toolbox)
- spmdSendReceive (Parallel Computing Toolbox)
- spmdSize (Parallel Computing Toolbox)
- spmdCat (Parallel Computing Toolbox)
- spmdReduce (Parallel Computing Toolbox)
- spmdPlus (Parallel Computing Toolbox)

These Image Processing Toolbox[™] functions now have thread-based support.

adapthisteq (Image	bfscore (Image	bwconncomp (Image	bwdist (Image
Processing Toolbox)	Processing Toolbox)	Processing Toolbox)	Processing Toolbox)
bwferet (Image	bwlabel (Image	bwlabeln (Image	bwlookup (Image
Processing Toolbox)	Processing Toolbox)	Processing Toolbox)	Processing Toolbox)
bwmorph3 (Image Processing Toolbox)	bwpack (Image Processing Toolbox)	bwskel (Image Processing Toolbox)	bwtraceboundary (Image Processing Toolbox)
bwunpack (Image	demosaic (Image	edge (Image Processing	edge3 (Image
Processing Toolbox)	Processing Toolbox)	Toolbox)	Processing Toolbox)
entropyfilt (Image Processing Toolbox)	exrHalfAsSingle (Image Processing Toolbox)	exrinfo (Image Processing Toolbox)	exrread (Image Processing Toolbox)
exrwrite (Image	fibermetric (Image	grabcut (Image	histeq (Image
Processing Toolbox)	Processing Toolbox)	Processing Toolbox)	Processing Toolbox)
hough (Image	imabsdiff (Image	<pre>imapplymatrix (Image Processing Toolbox)</pre>	imbothat (Image
Processing Toolbox)	Processing Toolbox)		Processing Toolbox)
imboxfilt (Image Processing Toolbox)	imboxfilt3 (Image Processing Toolbox)	imclose (Image Processing Toolbox)	imdiffusefilt (Image Processing Toolbox)
imdilate (Image	imerode (Image	imfill (Image	imfilter (Image
Processing Toolbox)	Processing Toolbox)	Processing Toolbox)	Processing Toolbox)

<pre>imgaussfilt (Image Processing Toolbox)</pre>	imgaussfilt3 (Image	imgradient (Image	imgradient3 (Image
	Processing Toolbox)	Processing Toolbox)	Processing Toolbox)
imhist (Image	imlincomb (Image	imnlmfilt (Image	imopen (Image
Processing Toolbox)	Processing Toolbox)	Processing Toolbox)	Processing Toolbox)
imreconstruct (Image Processing Toolbox)	imregionalmax (Image Processing Toolbox)	imregionalmin (Image Processing Toolbox)	imsegfmm (Image Processing Toolbox)
imtophat (Image Processing Toolbox)	inpaintCoherent (Image Processing Toolbox)	inpaintExemplar (Image Processing Toolbox)	integralImage (Image Processing Toolbox)
integralImage3 (Image Processing Toolbox)	intlut (Image Processing Toolbox)	iradon (Image Processing Toolbox)	isexr (Image Processing Toolbox)
label2idx (Image	medfilt2 (Image	medfilt3 (Image	modefilt (Image
Processing Toolbox)	Processing Toolbox)	Processing Toolbox)	Processing Toolbox)
multissim (Image	multissim3 (Image	nitfread (Image	obliqueslice (Image
Processing Toolbox)	Processing Toolbox)	Processing Toolbox)	Processing Toolbox)
ordfilt2 (Image	planar2raw (Image	poly2mask (Image	qtdecomp (Image
Processing Toolbox)	Processing Toolbox)	Processing Toolbox)	Processing Toolbox)
radon (Image	raw2planar (Image	raw2rgb (Image	rawread (Image
Processing Toolbox)	Processing Toolbox)	Processing Toolbox)	Processing Toolbox)
regionprops (Image Processing Toolbox)	regionprops3 (Image Processing Toolbox)	rgb2lab (Image Processing Toolbox)	rgb2lightness (Image Processing Toolbox)
superpixels (Image	superpixels3 (Image	watershed (Image	
Processing Toolbox)	Processing Toolbox)	Processing Toolbox)	

For more information, see Run MATLAB Functions in Thread-Based Environment.

GPU Functionality: Use functions with new and enhanced gpuArray support

These MATLAB functions have new and enhanced gpuArray (Parallel Computing Toolbox) support:

- allfinite (MATLAB)
- anymissing (MATLAB)
- anynan (MATLAB)
- ind2rgb (MATLAB)
- islocalmax (MATLAB)
- islocalmin (MATLAB)
- linsolve (MATLAB)
- mape (MATLAB)
- pagefun (Parallel Computing Toolbox) Support for norm (MATLAB)
- pagemldivide (MATLAB)

- pagemrdivide (MATLAB)
- pagenorm (MATLAB)
- rmoutliers (MATLAB) Support for defining outlier locations and returning the outlier indicator, thresholds, and center value outputs
- rmse (MATLAB)

For more information, see Run MATLAB Functions on a GPU (Parallel Computing Toolbox).

These Statistics and Machine Learning Toolbox functions have enhanced qpuArray support:

- fitcecoc (Statistics and Machine Learning Toolbox) Support for categorical predictors for classification tree learners
- fitcensemble (Statistics and Machine Learning Toolbox) Support for categorical predictors for classification tree learners
- fitcsvm (Statistics and Machine Learning Toolbox)
 - Support for hyperparameter optimization
 - When fitcsvm fits the SVM model on a GPU, some of the properties of the ClassificationSVM (Statistics and Machine Learning Toolbox) model are stored in GPU memory
 - Use gather (Parallel Computing Toolbox) to create a ClassificationSVM or CompactClassificationSVM object with properties stored in the local workspace from an equivalent object with properties stored in GPU memory
- fitctree (Statistics and Machine Learning Toolbox) Support for categorical predictors
- fitensemble (Statistics and Machine Learning Toolbox) Support for categorical predictors
- fitrensemble (Statistics and Machine Learning Toolbox) Support for categorical predictors
- fitrtree (Statistics and Machine Learning Toolbox) Support for categorical predictors
- pca (Statistics and Machine Learning Toolbox) Improved performance when you use the SVD algorithm (Parallel Computing Toolbox)
- Object functions of the CompactClassificationSVM (Statistics and Machine Learning Toolbox) model execute on a GPU if the model is fitted with GPU arrays
- Object functions of the ClassificationSVM (Statistics and Machine Learning Toolbox) model execute on a GPU if the model is fitted with GPU arrays

For a list of all Statistics and Machine Learning Toolbox functions with GPU functionality, see Functions with gpuArray support (Statistics and Machine Learning Toolbox).

These Signal Processing Toolbox functions have new and enhanced gpuArray support:

- instbw (Signal Processing Toolbox)
- instfreq (Signal Processing Toolbox)
- meanfreq (Signal Processing Toolbox)
- poctave (Signal Processing Toolbox)
- sosfilt (Signal Processing Toolbox) Support for FIR second-order subsections of the input filter

For a list of all Signal Processing Toolbox functions with GPU functionality, see Functions with gpuArray support (Signal Processing Toolbox).

This Audio Toolbox™ function has enhanced gpuArray support :

• audioFeatureExtractor (Audio Toolbox) - Code acceleration support for zerocrossrate

For a list of all Audio Toolbox functions with GPU functionality, see Functions with gpuArray support (Audio Toolbox).

This Image Processing Toolbox function has new gpuArray support:

• gray2ind (Image Processing Toolbox)

For a list of all Image Processing Toolbox functions with GPU functionality, see Functions with gpuArray support (Image Processing Toolbox).

Tall Arrays: Use functions with new and enhanced tall array support

These functions have new and enhanced tall array support:

- allfinite (MATLAB)
- anymissing (MATLAB)
- anynan (MATLAB)
- rmoutliers (MATLAB) Support for defining outlier locations and returning the outlier indicator, thresholds, and center value outputs
- · round (MATLAB) Support for specifying a direction for rounding ties

For more information, see Tall Arrays (MATLAB).

Distributed Arrays: Use functions with new and enhanced distributed array support

These functions have new and enhanced distributed array support:

- allfinite (MATLAB)
- anymissing (MATLAB)
- anynan (MATLAB)
- ind2rgb (MATLAB)
- mape (MATLAB)
- pageinv (MATLAB)
- pagemldivide (MATLAB)
- pagemrdivide (MATLAB)
- pagenorm (MATLAB)
- rank (MATLAB)
- rms (MATLAB)
- rmse (MATLAB)
- round (MATLAB) Support for specifying a direction for breaking ties

For more information, see Run MATLAB Functions with Distributed Arrays (Parallel Computing Toolbox).

MATLAB Job Scheduler: Support for your own Java installation

You can now use your own Java® 8 installation with MATLAB Job Scheduler (MJS). To use your own Java 8 installation with MJS, specify a path to a Java Runtime Environment (JRE) installation for the MJS_Java variable in your mjs_def file. For more information on where to find the mjs_def file, see Customize Startup Parameters.

Element-Wise Operations: Improved performance on GPU

Element-wise operations on large numbers of <code>gpuArray</code> objects show improved performance. Improvements are greater when you operate on large numbers of <code>gpuArray</code> objects. For example, adding one to every element of a cell array of the <code>gpuArray</code> data in this function is about 43.8x faster than in the previous release:

```
function timeElementWiseOps
```

```
% Prepare a cell array of gpuArray data
x = cell(5000,1);
x = cellfun(@(z) rand(10,"gpuArray"),x,"UniformOutput",false);
% Make all arrays different to one another
x = cellfun(@(z) z+rand(1,"gpuArray"),x,"UniformOutput",false);
% Time adding one to each element
gputimeit(@() cellfun(@(z) z+1,x,"UniformOutput",false))
```

The approximate execution times are:

R2022a: 5.70 seconds **R2022b:** 0.13 seconds

The code was timed on a Windows 10, Intel® Xeon® W-2133 @ 3.60 GHz test system with an NVIDIA RTX A5000 GPU by calling the timeElementWiseOps function.

svd Function: Improved performance on GPU

The svd (MATLAB) function shows improved performance when called with a gpuArray input matrix that is short and wide (n > 3m) or tall and narrow (m > 6n). For example, performing a singular value decomposition of a matrix in this function is about 1.4x faster than in the previous release:

```
function timeSVD
% Prepare input matrix
A = rand(1000,10000,"gpuArray");
% Time SVD
gputimeit(@() svd(A),3)
end
```

The approximate execution times are:

R2022a: 2.64 seconds

R2022b: 1.88 seconds

The code was timed on a Windows 10, Intel Xeon W-2133 @ 3.60 GHz test system with an NVIDIA RTX A5000 GPU by calling the time SVD function.

Functionality being removed or changed

labXxx functions are not recommended and are renamed to spmdXxx Still runs

To align these function names with their intended use within spmd blocks, these spmd block code execution and communication functions have been renamed:

- labBarrier (Parallel Computing Toolbox)
- labBroadcast (Parallel Computing Toolbox)
- labindex (Parallel Computing Toolbox)
- labProbe (Parallel Computing Toolbox)
- labReceive (Parallel Computing Toolbox)
- labSend (Parallel Computing Toolbox)
- labSendReceive (Parallel Computing Toolbox)
- numlabs (Parallel Computing Toolbox)

These functions are no longer recommended, but they will continue to work. This table shows the recommended replacement functions.

Functionality	Recommended Replacement	Compatibility Considerations
labBarrier	spmdBarrier (Parallel Computing Toolbox)	Replace all instances of labBarrier with spmdBarrier
labBroadcast	spmdBroadcast (Parallel Computing Toolbox)	Replace all instances of labBroadcast with spmdBroadcast
labindex	spmdIndex (Parallel Computing Toolbox)	Replace all instances of labindex with spmdIndex
labProbe	spmdProbe (Parallel Computing Toolbox)	Replace all instances of labProbe with spmdProbe
labReceive	spmdReceive (Parallel Computing Toolbox)	Replace all instances of labReceive with spmdReceive
labSend	spmdSend (Parallel Computing Toolbox)	Replace all instances of labSend with spmdSend

Functionality	Recommended Replacement	Compatibility Considerations
labSendReceive	spmdSendReceive (Parallel Computing Toolbox)	Replace all instances of labSendReceive with spmdSendReceive
numlabs	spmdSize (Parallel Computing Toolbox)	Replace all instances of numlabs with spmdSize

The labXxx functions will not be removed.

gcat, gop, and gplus are not recommended and are renamed to spmdCat, spmdReduce, and spmdPlus

Still runs

For performing SPMD operations on all the workers in an spmd block, gcat (Parallel Computing Toolbox), gop (Parallel Computing Toolbox), and gplus (Parallel Computing Toolbox) are no longer recommended. Use spmdCat (Parallel Computing Toolbox), spmdReduce (Parallel Computing Toolbox), and spmdPlus (Parallel Computing Toolbox) as direct replacements. The gcat, gop, and gplus functions will not be removed.

local profile has been renamed to Processes on the Parallel menu $Still\ runs$

For a process-based parallel environment on a local machine, the local profile is no longer recommended. Use Processes instead.

• To start a parallel pool of process workers, use this code.

```
parpool("Processes")
```

In previous releases, you used this code which is no longer recommended.

```
parpool("local")
```

The local profile option has been removed from the **Parallel** menu but will continue to work when you use it programmatically.

parallel.defaultClusterProfile and parallel.clusterProfiles are renamed to parallel.defaultProfile and parallel.listProfiles

Still runs

The parallel.defaultClusterProfile (Parallel Computing Toolbox) and parallel.clusterProfiles (Parallel Computing Toolbox) functions are renamed. parallel.defaultClusterProfile is renamed to parallel.defaultProfile (Parallel Computing Toolbox). parallel.clusterProfiles is renamed to parallel.listProfiles (Parallel Computing Toolbox). The parallel.defaultClusterProfile and parallel.clusterProfiles will not be removed.

To update your code, replace all instances of parallel.defaultClusterProfile with parallel.defaultProfile and parallel.clusterProfiles with parallel.listProfiles.

remotecopy has been removed

Errors

Starting in R2022b, remotecopy has been removed. To copy files to and from a remote host, use scp or sftp instead.

• Previously you used remotecopy and -protocol scp to copy files to and from a remote host using the secure copy protocol (SCP).

This table show to use scp instead.

Errors	Recommended	
remotecopy -remotehost host1 -local /my/	fsiclpe://prayt/rfi-ltec/partelmontoest/lr:e/moetnec/tfei/lfei/lpae/tphat	hprotocol s
	lsc/mmy/mfyi/lfei/lpee/tphathtohosntelmo/treem/onteem/oftiel/ef/ipleet, scp /my/file/path host2:/remote/file/pat	
remotecopy -remotehost host1 -local /my/	fsiclpe/postthl:-/fireomoter/efnictee/p/arte/mo/trey//ffiitlee//postth	h -protoco [·]

Previously you used remotecopy and -protocol sftp to copy files to and from a remote host using the secure file transfer protocol (SFTP).

This table shows how to use sftp instead.

Errors	Recommended	
remotecopy -remotehost host1 -local /my/	fsifttep/p/antyl/f-ittloe/praetnhotheos/trilen/orteen/oftiel/ef/iplaet/hpa	tphrotocol:
remotecopy -remotehost host1,host2 -loca	lsf/mpy//fmiyl/ef/iplæt/hpa-tthohorsemilot/ere/moetmec/tfei/lfei/lpea sftp/my/file/pathhost2:/remote/file/pa	
remotecopy -remotehost host1 -local /my/	fsifttep/phaots/t 1-:f/rroemo-tree/nfoitlee//praetnho t/en/yf/iftiel/ep/aptal	th-protoco

remotemjs has been removed

Errors

Starting in R2022b, remotemjs has been removed. Use ssh instead.

Previously you used remotemjs to run MATLAB Job Scheduler commands on a remote host using - protocol ssh or -protocol winsc.

This table shows how to use ssh instead.

Errors					Recommended		
remotemjs	<mjs< td=""><td>options></td><td>-matlabroot</td><td><instal< td=""><td>៤៩៩៧៨៣៩ជា៩ជា stenid feidernhoms៦វេ toolbox/paral</td><td>lel/bi</td><td>∟n/mj</td></instal<></td></mjs<>	options>	-matlabroot	<instal< td=""><td>៤៩៩៧៨៣៩ជា៩ជា stenid feidernhoms៦វេ toolbox/paral</td><td>lel/bi</td><td>∟n/mj</td></instal<>	៤៩៩៧៨៣៩ជា៩ជា stenid feidernhoms ៦វេ toolbox/paral	lel/bi	∟n/mj
remotemjs	<mjs< td=""><td>options></td><td>-matlabroot</td><td></td><td>ხნსსსონობლა nstemidtekidername და ქლიამაბა/paral ssh host2 <installfoldername>/toolbox/paral</installfoldername></td><td></td><td></td></mjs<>	options>	-matlabroot		ხნსსსონობლა nstemidtekidername და ქლიამაბა/paral ssh host2 <installfoldername>/toolbox/paral</installfoldername>		

R2022a

Version: 7.6

New Features

Compatibility Considerations

ValueStore and FileStore objects: Retrieve data and files on MATLAB clients during job execution

ValueStore (Parallel Computing Toolbox) and FileStore (Parallel Computing Toolbox) objects now allow you to store data and files from MATLAB workers that can be retrieved by MATLAB clients during the execution of a job (even while the job is still running). These objects are not held in system memory, so they can be used to store large results. A FileStore object provides a universal location for workers to copy files. MATLAB clients can then access these files regardless of the specific cluster environment where the worker is running.

The ValueStore and FileStore objects are automatically created when you create a job on a cluster, a parallel pool of process workers on your local machine, or a parallel pool of workers on a cluster of machines. To access the ValueStore and FileStore objects on a worker, use the getCurrentValueStore (Parallel Computing Toolbox) and getCurrentFileStore (Parallel Computing Toolbox) functions, respectively.

Multifactor Authentication for the Generic Scheduler Interface: Connect to remote clients with RemoteClusterAccess using two or more authentication factors

RemoteClusterAccess (Parallel Computing Toolbox) objects now allow you to perform multifactor authentication, including two-factor authentication. To do so, specify 'Multifactor' for the AuthenticationMode name-value argument. For details, see the RemoteClusterAccess reference page. The sample plugin scripts for third-party schedulers now accept 'Multifactor' as the value for the AuthenticationMode property of AdditionalProperties, which sets that value on RemoteClusterAccess.

Cluster resizing: Customize your MATLAB Job Scheduler cluster to resize automatically

You can customize your MATLAB Job Scheduler (MJS) cluster to resize automatically. After you set up auto-resizing, also called auto-scaling, the cluster can automatically change the number of workers with the amount of work submitted. The cluster grows (scales up) when there is more work to do and shrinks (scales down) when there is less work to do. This allows you to use your compute resources more efficiently and can result in cost savings. To learn more, see Set up MATLAB Job Scheduler Cluster for Auto-Resizing.

Parallel Pools: Check status of pools

Starting in R2022a, you can query if any type of pool is currently running by using the Busy property of the pool. This property indicates whether the parallel pool is busy, specified as true or false. The pool is busy if there is outstanding work for the pool to complete.

Background Pool: Number of thread workers no longer limited to 8 workers

Before R2022a, the NumWorkers property was capped at a maximum of 8 workers when Parallel Computing Toolbox was installed, and 1 worker when not installed. This limit is no longer in place

when the Parallel Computing Toolbox is installed. When Parallel Computing Toolbox is not installed, the background pool remains limited to 1 worker.

Thread-Based Parallel Pool: See futures in active pool

Starting in R2022a, you can query all queued and running futures on a parallel pool by using the FevalQueue property of the pool. To create futures, use parfeval (Parallel Computing Toolbox) and parfevalOnAll (Parallel Computing Toolbox). For more information on futures, see Future (MATLAB).

cancelAll Method: Cancel currently queued and running futures in a parallel pool

cancelAll (Parallel Computing Toolbox) cancels all futures currently queued or running in a parallel pool. Queued or running futures are listed in the FevalQueue property.

For example, you can use cancelAll to stop all Futures in an FevalQueue.

```
pool = parpool;
cancelAll(pool.FevalQueue);
```

GPU Functionality: Use new and enhanced gpuArray functions

New GPU support in MATLAB:

- besseli (MATLAB), besselk (MATLAB)
- besselj (MATLAB), bessely (MATLAB) Support for scale input argument
- deconv (MATLAB)
- eps (MATLAB), flintmax (MATLAB), intmax (MATLAB), intmin (MATLAB), realmax (MATLAB), realmin (MATLAB) — Support for "like" syntax to return scalars from prototype object
- filter (MATLAB) Support for all syntaxes and any vector sizes of the numerator and denominator coefficients of a rational transfer function in the input arguments
- pagefun (Parallel Computing Toolbox) Support for conv2 (MATLAB), qr (MATLAB), and svd (MATLAB)
- pageinv (MATLAB)
- pagesvd (MATLAB)
- svd (MATLAB)

For more information, see Run MATLAB Functions on a GPU (Parallel Computing Toolbox).

New GPU support in Statistics and Machine Learning Toolbox:

- fitensemble (Statistics and Machine Learning Toolbox)
- fitrensemble (Statistics and Machine Learning Toolbox)
- New support for probability functions: gev*, gp*, nbin*

For a list of all Statistics and Machine Learning Toolbox functions with GPU functionality, see Functions with gpuArray support (Statistics and Machine Learning Toolbox).

The following functions have new and enhanced gpuArray support in Signal Processing Toolbox:

- findpeaks (Signal Processing Toolbox)
- zerocrossrate (Signal Processing Toolbox)
- pkurtosis (Signal Processing Toolbox)
- pentropy (Signal Processing Toolbox)
- levinson (Signal Processing Toolbox)
- hilbert (Signal Processing Toolbox)
- pow2db (Signal Processing Toolbox)
- db2pow (Signal Processing Toolbox)

For a list of all Signal Processing Toolbox functions with GPU functionality, see Functions with gpuArray support (Signal Processing Toolbox).

The following functions have new gpuArray support in Audio Toolbox:

- pitch (Audio Toolbox)
- audioDataAugmenter (Audio Toolbox)
- audioFeatureExtractor (Audio Toolbox) improved support

For a list of all Audio Toolbox functions with GPU functionality, see Functions with gpuArray support (Audio Toolbox).

The following functions have new gpuArray and dlArray support in Wavelet Toolbox:

- imodwt (Wavelet Toolbox)
- modwt (Wavelet Toolbox)
- modwtmra (Wavelet Toolbox)

For a list of all Wavelet Toolbox functions with GPU functionality, see Functions with gpuArray support (Wavelet Toolbox).

Support for NVIDIA CUDA 11.2: Update to CUDA Toolkit 11.2

The parallel computing products now use CUDA toolkit version 11.2. To generate CUDA kernel objects from CU code or compile CUDA compatible source code, libraries, and executables using GPU Coder™, you must use toolkit version 11.2. For more information, see Run CUDA or PTX Code on GPU (Parallel Computing Toolbox).

Tall Arrays: Use new and enhanced tall array functionality

 eps (MATLAB), flintmax (MATLAB), intmax (MATLAB), intmin (MATLAB), realmax (MATLAB), realmin (MATLAB) — Support for "like" syntax to return scalars from prototype object

For more information, see Tall Arrays (MATLAB).

Distributed Arrays: Use new and enhanced distributed array functionality

- deconv (MATLAB)
- eps (MATLAB), flintmax (MATLAB), intmax (MATLAB), intmin (MATLAB), realmax (MATLAB), realmin (MATLAB) Support for "like" syntax to return scalars from prototype object
- interp3 (MATLAB), interpn (MATLAB)
- movmad (MATLAB), movmax (MATLAB), movmean (MATLAB), movmedian (MATLAB), movmin (MATLAB), movprod (MATLAB), movstd (MATLAB), movsum (MATLAB), movvar (MATLAB)

For more information, see Run MATLAB Functions with Distributed Arrays (Parallel Computing Toolbox).

Parallel Features in Other Products

Parallel features added in other products:

- Experiment Manager: Offload deep learning experiments as batch jobs in a cluster
 - Starting in R2022a, **Experiment Manager** (Deep Learning Toolbox) supports offloading experiments as batch jobs in a cluster. You can configure the cluster to run multiple trials at the same time or to run a single trial at a time on multiple parallel workers. While the experiment is running in the cluster, you can run other experiments, close the app and continue using MATLAB, or close your MATLAB session. For more information, see Offload Experiments as Batch Jobs to Cluster (Deep Learning Toolbox).
- Parallel Simulations: Perform parameter sweeps using Parameter Combinations
 - In R2022a, you can use Parameter Combinations in the Multiple Simulations panel of the Simulink $^{\otimes}$ Editor for workflows with multiple simulations, such as Monte-Carlo simulations and parameter sweeps. Parameter Combinations allows you to create sequential and exhaustive combinations of parameters, specify value ranges, and run simulations with these combinations. The Multiple Simulations panel was introduced in R2021b.
- Machine Learning Apps: Train draft models in parallel or train in the background to keep the app responsive
 - In Classification Learner (Statistics and Machine Learning Toolbox) and Regression Learner (Statistics and Machine Learning Toolbox), you can train multiple draft models in parallel. You can use the **Use Parallel** or **Use Background** buttons while training models.

Solving Linear System: Improved performance when solving linear systems A*X = B with gpuArray for symmetric positive definite matrices A

Solving a linear system of the form A*X = B with gpuArray by executing $A\B$ shows improved performance when A is a symmetric positive definite matrix.

For example, this code solves A*X = B for a 10,000-by-10,000 symmetric positive definite matrix A and a 10,000-by-1 column vector B. The code is about 2.4x faster than in the previous release.

```
function timingTest
rng default
R = rand(10000, "gpuArray");
A = R'*R;
B = ones(10000,1, "gpuArray");
X = A\B;
end
```

The approximate execution times are:

R2021b: 1.03 s **R2022a:** 0.43 s

The code was timed on a Windows 10, Intel Xeon CPU E5-1640 v3 @ 3.50 GHz with an NVIDIA Titan V GPU test system using the gputimeit function:

gputimeit(@timingTest)

Functionality being removed or changed

Support for macOS has been removed

Errors

Support for macOS has been removed in R2022a. To run the MATLAB Parallel Server cluster, use a Windows or Linux® operating system instead.

You can still connect to a MATLAB Parallel Server cluster when you use Parallel Computing Toolbox running on a macOS client.

distcomp folder removed, now named parallel

The distcomp folder has been removed.

In R2019b, the Parallel Computing Toolbox folder was renamed. Since then, the name of the toolbox folder has been parallel. If you need to reference the location of the toolbox, update your references to use toolbox/parallel instead of toolbox/distcomp.

R2021b

Version: 7.5

New Features

Compatibility Considerations

Parallel Language in MATLAB: Share parallel code with any MATLAB user

From R2021b, you can use more parallel language features in serial without Parallel Computing Toolbox. To scale up and speed up computations that use these features, use parallel pools and Parallel Computing Toolbox.

Additionally, you can now share parallel code with other MATLAB users who do not have Parallel Computing Toolbox.

The following features are now available in MATLAB:

- parfeval and related functionality such as afterEach and afterAll
- parallel.pool.DataQueue, parallel.pool.PollableDataQueue, and related functionality such as afterEach

For more information, see Background Processing (MATLAB) and Write Portable Parallel Code (Parallel Computing Toolbox).

GPU Functionality: Use new and enhanced gpuArray functions

- makima (MATLAB)
- movmad (MATLAB)
- movmax (MATLAB)
- movmedian (MATLAB)
- movmin (MATLAB)
- movprod (MATLAB)
- pchip (MATLAB)
- ppval (MATLAB)
- filter (MATLAB) Support for more than 10 elements for numerator coefficients of rational transfer function input argument.
- pagefun (Parallel Computing Toolbox) Support for conv (MATLAB)

For more information, see Run MATLAB Functions on a GPU (Parallel Computing Toolbox).

GPU Functionality: Use new and enhanced gpuArray functions in Statistics and Machine Learning Toolbox

- betafit (Statistics and Machine Learning Toolbox)
- fitcecoc (Statistics and Machine Learning Toolbox)
- fitcensemble (Statistics and Machine Learning Toolbox)
- fitctree (Statistics and Machine Learning Toolbox)
- fitdist (Statistics and Machine Learning Toolbox)
- fitrtree (Statistics and Machine Learning Toolbox)
- gevfit (Statistics and Machine Learning Toolbox)

- gpfit (Statistics and Machine Learning Toolbox)
- ksdensity (Statistics and Machine Learning Toolbox)
- mle (Statistics and Machine Learning Toolbox)
- mvksdensity (Statistics and Machine Learning Toolbox)
- nbinfit (Statistics and Machine Learning Toolbox)

For a list of all Statistics and Machine Learning Toolbox functions with GPU functionality, see Functions with gpuArray support (Statistics and Machine Learning Toolbox).

GPU Functionality: Use new and enhanced gpuArray functions for working with signals and audio

- The following functions have new and enhanced gpuArray support in Signal Processing Toolbox: For a list of all Signal Processing Toolbox functions with GPU functionality, see Functions with gpuArray support (Signal Processing Toolbox).
 - downsample (Signal Processing Toolbox)
 - filtfilt (Signal Processing Toolbox)
 - pspectrum (Signal Processing Toolbox)
 - resample (Signal Processing Toolbox)
 - shiftdata (Signal Processing Toolbox)
 - unshiftdata (Signal Processing Toolbox)
 - upfirdn (Signal Processing Toolbox)
 - upsample (Signal Processing Toolbox)
 - xspectrogram (Signal Processing Toolbox)
- The following functions have new gpuArray support in Audio Toolbox:
 - classifySound (Audio Toolbox)
 - crepePostprocess (Audio Toolbox)
 - crepePreprocess (Audio Toolbox)
 - detectSpeech (Audio Toolbox)
 - harmonicRatio (Audio Toolbox)
 - openl3Features (Audio Toolbox)
 - openl3Preprocess (Audio Toolbox)
 - pitchnn (Audio Toolbox)
 - vggishFeatures (Audio Toolbox)
 - vggishPreprocess (Audio Toolbox)
 - yamnetPreprocess (Audio Toolbox)

For a list of all Audio Toolbox functions with GPU functionality, see Functions with gpuArray support (Audio Toolbox).

Memory Usage: Use whos to check memory used by gpuArray and distributed variables

You can now use the whos (MATLAB) function to check the amount of memory used by gpuArray and distributed variables.

Previously, the Bytes variable of the output of whos showed the number of bytes of the pointer to the variable in the local memory of the host machine. Now, the Bytes variable displays the amount of GPU or distributed memory allocated to that variable. For gpuArray variables, whos displays the amount of GPU memory used by that variable. For distributed variables, whos displays the total memory used by that variable across all workers in the pool.

Distributed Arrays: Use new and enhanced distributed array functionality

- interp2 (MATLAB)
- linsolve (MATLAB)
- makima (MATLAB)
- orth (MATLAB)
- pagesvd (MATLAB)
- pchip (MATLAB)
- subspace (MATLAB)
- svdsketch (MATLAB)
- decomposition (MATLAB) Support for new decompositions
 - For dense matrices, new support for banded decomposition and permuted triangular decomposition
 - For sparse matrices, new support for LDL decomposition and QR decomposition
- pagefun (Parallel Computing Toolbox) Support for conv (MATLAB), conv2 (MATLAB), and svd (MATLAB)

For more information, see Run MATLAB Functions with Distributed Arrays (Parallel Computing Toolbox).

Thread-Based Environment: Use new and enhanced functionality on threads for working with audio, video, and images

- The following MATLAB functions have new and enhanced thread support:
 - imresize (MATLAB)
 - audioread (MATLAB)
 - audiowrite (MATLAB)
 - imwrite (MATLAB)
 - VideoReader (MATLAB)
 - VideoWriter (MATLAB)

- imread (MATLAB) Support for JPEG 2000 (J2C, J2K, JP2, JPF, JPX) format images on threads
- The following functions and objects have new thread support in Image Processing Toolbox:
 - imcrop (Image Processing Toolbox)
 - imcrop3 (Image Processing Toolbox)
 - imresize3 (Image Processing Toolbox)
 - imrotate (Image Processing Toolbox)
 - imrotate3 (Image Processing Toolbox)
 - imtranslate (Image Processing Toolbox)
 - impyramid (Image Processing Toolbox)
 - imwarp (Image Processing Toolbox)
 - affineOutputView (Image Processing Toolbox)
 - findbounds (Image Processing Toolbox)
 - fliptform (Image Processing Toolbox)
 - makeresampler (Image Processing Toolbox)
 - maketform (Image Processing Toolbox)
 - tformarray (Image Processing Toolbox)
 - tformfwd (Image Processing Toolbox)
 - tforminv (Image Processing Toolbox)
 - imregconfig (Image Processing Toolbox)
 - imregcorr (Image Processing Toolbox)
 - imregdemons (Image Processing Toolbox)
 - imregmtb (Image Processing Toolbox)
 - normxcorr2 (Image Processing Toolbox)
 - MattesMutualInformation (Image Processing Toolbox)
 - MeanSquares (Image Processing Toolbox)
 - RegularStepGradientDescent (Image Processing Toolbox)
 - OnePlusOneEvolutionary (Image Processing Toolbox)
 - cpcorr (Image Processing Toolbox)
 - cpstruct2pairs (Image Processing Toolbox)
 - fitgeotrans (Image Processing Toolbox)
 - imref2d (Image Processing Toolbox)
 - imref3d (Image Processing Toolbox)
 - affine2d (Image Processing Toolbox)
 - affine3d (Image Processing Toolbox)
 - projective2d (Image Processing Toolbox)
 - gray2ind (Image Processing Toolbox)
 - ind2gray (Image Processing Toolbox)
 - mat2gray (Image Processing Toolbox)
 - label2rgb (Image Processing Toolbox)

- imsplit (Image Processing Toolbox)
- adaptthresh (Image Processing Toolbox)
- otsuthresh (Image Processing Toolbox)
- imquantize (Image Processing Toolbox)
- grayslice (Image Processing Toolbox)
- im2int16 (Image Processing Toolbox)
- im2single (Image Processing Toolbox)
- im2uint16 (Image Processing Toolbox)
- im2uint8 (Image Processing Toolbox)

For more information, see Run MATLAB Functions in Thread-Based Environment (MATLAB).

Reference Architectures: Deploy and Run MATLAB Parallel Server from Azure Marketplace

Quickly and easily deploy MATLAB Parallel Server in the cloud using bring-your-own-license (BYOL) software plans developed by MathWorks® for Azure® Marketplace.

The MATLAB Parallel Server BYOL reference architecture for Azure Marketplace now supports R2021a. The MATLAB Parallel Server BYOL reference architecture was released in February 2021 with support for R2020b.

For more information, see Run MATLAB Parallel Server from Azure Marketplace.

Reference Architectures: Deploy and Run Network License Manager from Azure Marketplace

Quickly and easily deploy a network license manager in the cloud using software plans developed by MathWorks for Azure Marketplace.

The network license manager reference architecture for Azure Marketplace now supports R2021a. The network license manager for Azure Marketplace was released in February 2021 with support for R2020b.

For more information, see Run Network License Manager from Azure Marketplace.

Functionality Being Removed or Changed

Support for macOS will be removed

Still runs

Support for macOS will be removed in R2022a. To run the MATLAB Parallel Server cluster, use a Windows or Linux operating system instead.

You can still connect to a MATLAB Parallel Server cluster when you use Parallel Computing Toolbox running on a macOS client.

parfeval and parfevalOnAll can now run in serial with no pool

Behavior change

Starting in R2021b, you can now run parfeval (Parallel Computing Toolbox) and parfevalOnAll (Parallel Computing Toolbox) in serial with no pool. This behavior allows you to share parallel code that you write with users who do not have Parallel Computing Toolbox.

When you use the syntaxes parfeval(fcn,n,X1,...,Xm) or parfevalOnAll(fcn,n,X1,...,Xm), MATLAB tries to use an open parallel pool if you have Parallel Computing Toolbox. If a parallel pool is not open, MATLAB will create one if automatic pool creation is enabled.

If parallel pool creation is disabled or if you do not have Parallel Computing Toolbox, the function is evaluated in serial. In previous releases, MATLAB threw an error instead.

mldivide and decomposition now produce the same results for distributed arrays Behavior change

Starting in R2021b, mldivide and decomposition now produce the same results when you run the following code using distributed arrays A and b.

```
X = A \ b;
X = decomposition(A) \ b;
```

Previously, you sometimes saw different results when you used decomposition before mldivide for a nonsquare distributed matrix A.

Default behavior for decomposition has changed for distributed arrays *Behavior change*

Starting in R2021b, the algorithm for decomposition with type set to 'auto' (default) has changed for distributed array input. You see this behavior change when you use decomposition to decompose a distributed matrix that is Hermitian, banded, or permuted triangular.

When you use decomposition(A) or decomposition(A, 'auto') with a distributed matrix A, the type of decomposition selected by MATLAB is limited to the supported decomposition types for distributed arrays.

- For a distributed dense matrix A, in the syntax decomposition (A, type) the decomposition types 'ldl', 'cod', and 'hessenberg' are not supported.
- For a distributed sparse matrix A, in the syntax decomposition (A, type) the decomposition types 'chol', 'cod', and 'hessenberg' are not supported.

remotecopy will be removed

Warns

remotecopy will be removed in a future release.

• Previously you used remotecopy and -protocol scp to copy files to and from a remote host using the scp protocol.

As a best practice use scp instead.

Not Recommended					Recommended		
remotecopy	-remotehost	host1	-local	/my/	fsictpe://prayt/rfi-ttec/partelmohtosst/1r.e/moetneo/tfei/Ufei/Upee/tpat	protoc	ol s
remotecopy	-remotehost	host1	,host2		lsc/mmy/mfyi/lfei/lpme/tphathtohosntelmo/treem/onteem/oftiel/ef/iplact, scp /my/file/path host2:/remote/file/pat		prot

Not Recommended	Recommended	
remotecopy -remotehost host1 -local /my/	fsiclpe/poestthl:-/fneomoter/efnicltee/p/arte/mg/tney//ffiillee//poestth	ı-protocol

• Previously you used remotecopy and -protocol sftp to copy files to and from a remote host using the sftp protocol.

As a best practice use sftp instead.

Not Recommended	Recommended	
remotecopy -remotehost host1 -local /my/	fsifttep/p/antyl/f-ittoe/praetnhotheos/trien/orteen/oftiel/ef/iplaet/hpa	tphrotocol:
remotecopy -remotehost host1,host2 -loca	lsf/mpy//fmiyl/ef/iplæt/hpatthohorsemilot/ere/moetmey/tfei/lfei/lpea sftp/my/file/pathhost2:/remote/file/pa	φath -pro th
remotecopy -remotehost host1 -local /my/	fsiftep/praotst: 1-:f/rroemo-tree/rifoitlee//praetrinot/en/yf/iftiel/ep/aptal	th-protoco

remotemjs will be removed

Warns

remotemjs will be removed in a future release.

Previously you used remotemjs to run MATLAB Job Scheduler commands on a remote host using - protocol ssh or -protocol winsc.

As a best practice use ssh instead.

Not Recommended	Recommended	
remotemjs <mjs options=""> -matlabroot <instal< td=""><td>bsbluestame≥instenidteNoetnamst∤toolbox/paral</td><td>lel/bin/mj</td></instal<></mjs>	b sbluesta me≥inst enidteNoetnams t∤toolbox/paral	lel/bin/mj
	ბნ ს∖სოცისპოლ ins temlotfeklokrhamgt∤ქსით tხ∆x/paral ssh host2 <i><installfoldername< i="">>/toolbox/paral</installfoldername<></i>	

R2021a

Version: 7.4

New Features

Compatibility Considerations

GPU Functionality: Use new and enhanced gpuArray functions

- pagectranspose (MATLAB)
- pagetranspose (MATLAB)
- spline (MATLAB)
- imresize (MATLAB) Support for all non-sparse numeric or logical images, except categorical or indexed images. Support for 'nearest' and 'bilinear' interpolation methods. Support for 'box', 'triangle', 'cubic', 'lanczos2', 'lanczos3', and custom interpolation kernels. Support for 'AntiAliasing' name-value argument.
- issorted (MATLAB) Support for dim and direction input arguments and ComparisonMethod name-value option.

For more information, see Run MATLAB Functions on a GPU (Parallel Computing Toolbox).

GPU Functionality: Use new and enhanced gpuArray functions in Statistics and Machine Learning Toolbox

- fitcensemble (Statistics and Machine Learning Toolbox)
- fitcknn (Statistics and Machine Learning Toolbox)
- gamfit (Statistics and Machine Learning Toolbox)
- pca (Statistics and Machine Learning Toolbox)
- randsample (Statistics and Machine Learning Toolbox)

For a list of all Statistics and Machine Learning Toolbox functions with GPU functionality, see Functions with gpuArray support (Statistics and Machine Learning Toolbox).

GPU Functionality: Use new and enhanced gpuArray functions for working with signals, audio, and wavelets

- The following functions have new and enhanced gpuArray support in Signal Processing Toolbox:
 - binmask2sigroi (Signal Processing Toolbox)
 - bitrevorder (Signal Processing Toolbox)
 - buffer (Signal Processing Toolbox)
 - digitrevorder (Signal Processing Toolbox)
 - extendsigroi (Signal Processing Toolbox)
 - mergesigroi (Signal Processing Toolbox)
 - removesigroi (Signal Processing Toolbox)
 - shortensigroi (Signal Processing Toolbox)
 - sigroi2binmask (Signal Processing Toolbox)
 - sosfilt (Signal Processing Toolbox)
 - stftmag2sig (Signal Processing Toolbox)

For a list of all Signal Processing Toolbox functions with GPU functionality, see Functions with gpuArray support (Signal Processing Toolbox).

- The following functions have new gpuArray support in Wavelet Toolbox:
 - appcoef (Wavelet Toolbox)
 - appcoef2 (Wavelet Toolbox)
 - detcoef (Wavelet Toolbox)
 - detcoef2 (Wavelet Toolbox)
 - haart (Wavelet Toolbox)
 - haart2 (Wavelet Toolbox)
 - idwt (Wavelet Toolbox)
 - idwt2 (Wavelet Toolbox)
 - ihaart2 (Wavelet Toolbox)
 - ihaart (Wavelet Toolbox)
 - waveletScattering (Wavelet Toolbox)
 - waverec (Wavelet Toolbox)
 - waverec2 (Wavelet Toolbox)

For a list of all Wavelet Toolbox functions with GPU functionality, see Functions with gpuArray support (Wavelet Toolbox).

- The following function has new gpuArray support in Audio Toolbox:
 - audioFeatureExtractor (Audio Toolbox)

For a list of all Audio Toolbox functions with GPU functionality, see Functions with gpuArray support (Audio Toolbox).

GPU Functionality: Use new and enhanced gpuArray functions in Image Processing Toolbox

- imwarp (Image Processing Toolbox)
- imcrop (Image Processing Toolbox)
- multissim (Image Processing Toolbox)
- multissim3 (Image Processing Toolbox)
- psnr (Image Processing Toolbox)
- wiener2 (Image Processing Toolbox)

For a list of all Image Processing Toolbox functions with GPU functionality, see Functions with gpuArray support (Image Processing Toolbox).

Support for NVIDIA CUDA 11.0: Update to CUDA Toolkit 11.0

The parallel computing products are now using CUDA toolkit version 11.0. To generate CUDA kernel objects from CU code or compile CUDA compatible source code, libraries, and executables using GPU Coder, you must use toolkit version 11.0. For more information, see GPU Support by Release (Parallel Computing Toolbox).

Distributed Arrays: Use new and enhanced distributed array functionality

- colon (MATLAB)
- ichol (MATLAB)
- interp1 (MATLAB)
- pagectranspose (MATLAB)
- pagemtimes (MATLAB)
- pagetranspose (MATLAB)

For more information, see Run MATLAB Functions with Distributed Arrays (Parallel Computing Toolbox).

GPU Devices: Count and compare available GPU devices

Use new functionality to count and compare GPU devices in your local system

- gpuDeviceCount (Parallel Computing Toolbox) Count all detected devices, all supported devices, or only devices available in your current session.
- gpuDeviceTable (Parallel Computing Toolbox) Compare the properties of all local GPU devices detected in your system.

Thread-Based Parallel Pool: Use new and enhanced functionality on thread workers

- imageDatastore (MATLAB)
- imread (MATLAB)
- parallel.pool.DataQueue (Parallel Computing Toolbox)
- parallel.pool.PollableDataQueue (Parallel Computing Toolbox)
- Use thread workers in standalone applications created with MATLAB Compiler $^{\text{\tiny TM}}$ and web apps hosted on MATLAB Web App Server $^{\text{\tiny TM}}$.

For more information, see Run MATLAB Functions on Thread Workers (Parallel Computing Toolbox).

parfor Examples: Use a parallel pool to speed up Monte-Carlo code

Use this new example to learn how to speed up your Monte-Carlo code using a parfor-loop. For more information, see Use parfor to Speed Up Monte-Carlo Code (Parallel Computing Toolbox).

Upgrade Parallel Computing Products Together

As with every new release, if you upgrade Parallel Computing Toolbox, you must upgrade MATLAB Parallel Server at the same time so that they interact properly with each other.

Compatibility Considerations

If you are using MATLAB Job Scheduler, the backward compatibility feature allows you to connect to multiple versions of MATLAB Parallel Server in your cluster. For more information, see Run Multiple MATLAB Parallel Server Versions.

Jobs created in one version of Parallel Computing Toolbox software do not run in a different version of MATLAB Parallel Server software, and might not be readable in different versions of the toolbox software. The job data stored in the folder identified by JobStorageLocation might not be compatible between different versions of MATLAB Parallel Server. You must specify a different JobStorageLocation for each parallel computing product, and each version on your cluster must have its own JobStorageLocation.

Functionality Being Removed or Changed

Support for Kepler GPUs (CC 3.0 and 3.2) is removed

Errors

Starting in R2021a, support for Kepler GPU architectures with compute capability 3.0 and 3.2 is removed. Using a GPU with MATLAB requires a GPU device with compute capability 3.5 or greater. Using a GPU device with compute capability 6.0 or greater is recommended.

For more information about supported GPU devices, see GPU Support by Release (Parallel Computing Toolbox).

Functions offloaded with batch now evaluate cell array input arguments $\{C1,...,Cn\}$ as C1,...,Cn

Behavior change

Starting in R2021a, a function fcn offloaded with batch (Parallel Computing Toolbox) evaluates cell array input arguments $\{C1, \ldots, Cn\}$ as fcn $\{C1, \ldots, Cn\}$. In previous releases $\{C1, \ldots, Cn\}$ threw an error and $\{\{C1, \ldots, Cn\}\}$ was evaluated as fcn $\{C1, \ldots, Cn\}$.

Starting in R2021a, use the following code to offload $fcn({a,b},{c,d})$ on the cluster myCluster with one output.

```
batch(myCluster,@fcn,1,{{a,b},{c,d}});
```

In previous releases, you used the following code instead.

```
batch(myCluster,@fcn,1,{{{a,b},{c,d}}});
```

R2020b

Version: 7.3

New Features

Compatibility Considerations

GPU Functionality: Use new and enhanced gpuArray functions

- isgpuarray (Parallel Computing Toolbox)
- isoutlier (MATLAB)
- rmmissing (MATLAB)
- rmoutliers (MATLAB)
- Message functions assert (MATLAB), error (MATLAB), and warning (MATLAB) (these functions gather gpuArray data and run on the CPU)
- arrayfun (Parallel Computing Toolbox) Support for cast (MATLAB) using 'like' syntax
- qr (MATLAB) Support for one- and three-argument syntaxes

For more information, see Run MATLAB Functions on a GPU (Parallel Computing Toolbox).

GPU Functionality: Use new and enhanced gpuArray functions in Statistics and Machine Learning Toolbox

- fitlm (Statistics and Machine Learning Toolbox)
- fitglm (Statistics and Machine Learning Toolbox)
- glmfit (Statistics and Machine Learning Toolbox)
- glmval (Statistics and Machine Learning Toolbox)
- grp2idx (Statistics and Machine Learning Toolbox)

For a list of all Statistics and Machine Learning Toolbox functions with GPU functionality, see Functions with gpuArray support (Statistics and Machine Learning Toolbox).

GPU Functionality: Use new and enhanced gpuArray functions for working with signals, audio, and wavelets

- The following functions have new and enhanced gpuArray support in Signal Processing Toolbox:
 - cpsd (Signal Processing Toolbox)
 - fsst (Signal Processing Toolbox)
 - goertzel (Signal Processing Toolbox)
 - istft (Signal Processing Toolbox)
 - mscohere (Signal Processing Toolbox)
 - periodogram (Signal Processing Toolbox)
 - pwelch (Signal Processing Toolbox)
 - spectrogram (Signal Processing Toolbox) Support for nonuniformly spaced frequencies

For a list of all Signal Processing Toolbox functions with GPU functionality, see Functions with gpuArray support (Signal Processing Toolbox).

- The following functions have new gpuArray support in Wavelet Toolbox:
 - dwt (Wavelet Toolbox)
 - dwt2 (Wavelet Toolbox)

- dyadup (Wavelet Toolbox)
- dyaddown (Wavelet Toolbox)
- timeSpectrum
- scaleSpectrum
- wavedec (Wavelet Toolbox)
- wavedec2 (Wavelet Toolbox)
- wcoherence (Wavelet Toolbox)
- wextend (Wavelet Toolbox)
- wkeep (Wavelet Toolbox)

For a list of all Wavelet Toolbox functions with GPU functionality, see Functions with gpuArray support (Wavelet Toolbox).

- The following functions have new gpuArray support in Audio Toolbox:
 - audioDelta (Audio Toolbox)
 - cepstralCoefficients (Audio Toolbox)
 - shiftPitch (Audio Toolbox)
 - spectralCentroid (Audio Toolbox)
 - spectralCrest (Audio Toolbox)
 - spectralDecrease (Audio Toolbox)
 - spectralEntropy (Audio Toolbox)
 - spectralFlatness (Audio Toolbox)
 - spectralFlux (Audio Toolbox)
 - spectralKurtosis (Audio Toolbox)
 - spectralRolloffPoint (Audio Toolbox)
 - spectralSkewness (Audio Toolbox)
 - spectralSlope (Audio Toolbox)
 - spectralSpread (Audio Toolbox)
 - stretchAudio (Audio Toolbox)

For a list of all Audio Toolbox functions with GPU functionality, see Functions with gpuArray support (Audio Toolbox).

GPU Communications: Fast data transfer between GPUs in a parallel pool

Data transfer between GPUs in a parallel pool now uses fast peer-to-peer communication, including NVLink, if available. Fast data transfer takes place when you send gpuArray data using the following functions:

- labSend (Parallel Computing Toolbox)
- labReceive (Parallel Computing Toolbox)
- labBroadcast (Parallel Computing Toolbox)

• gop (Parallel Computing Toolbox)

Support for NVIDIA CUDA 10.2: Update to CUDA Toolkit 10.2

The parallel computing products are now using CUDA toolkit version 10.2. To compile CUDA code for CUDAKernel or CUDA MEX-files, you must use toolkit version 10.2. For more information, see GPU Support by Release (Parallel Computing Toolbox).

Distributed Arrays: Use new and enhanced distributed array functionality

- filter (MATLAB)
- mpower (MATLAB)
- pagefun (Parallel Computing Toolbox)
- rmmissing (MATLAB)

For more information, see Run MATLAB Functions with Distributed Arrays (Parallel Computing Toolbox).

Tall Arrays: Use new and enhanced tall array functionality in Signal Processing Toolbox

pwelch (Signal Processing Toolbox)

For a list of all Signal Processing Toolbox functions with support for tall arrays, see Functions with tall support (Signal Processing Toolbox).

Array Assignment: Assign gpuArray or distributed data directly into existing MATLAB arrays

You can now assign values stored in a gpuArray, distributed array, or codistributed array into a MATLAB array directly, without first having to gather the data. In assignment statements such as A(1:k) = C, where A has a built-in data type such as double, MATLAB attempts to convert C to the same data type as A. That conversion behavior has changed.

For example, the following code now runs.

```
x = rand(3,3);
y = rand(1,'gpuArray');
x(2,2) = y;
```

Compatibility Considerations

Some assignment statements that used to throw an error now execute. If your code relied on the errors that MATLAB threw for those conversions, such as within a try/catch block, then your code might no longer catch those errors.

Distributed Tables: Assign data directly into existing distributed tables

You can now assign values directly into a distributed table. In assignment statements such as distTable.Score(1:k) = C, where distTable.Score has a built-in data type such as double, MATLAB attempts to convert C to be the same data type as distTable.Score. That conversion behavior has changed.

For example, the following code now runs.

```
table = array2table(rand(2), 'VariableNames', {'ID', 'Score'});
distTable = distributed(table);
distTable.Score(1:2) = [1 2];
```

Compatibility Considerations

Some assignment statements that used to throw an error now execute. If your code relied on the errors that MATLAB threw when assigning values into a distributed table, such as within a try/catch block, then your code might no longer catch those errors.

Thread-Based Parallel Pool: Use a gpuArray on thread workers

You can now use a gpuArray on thread workers. Any functionality with both GPU and ThreadPool support now supports GPUs in a ThreadPool. For more information, see Check Support for Thread-Based Environment (Parallel Computing Toolbox).

parfeval Examples: Explore the state of futures and cancel them

Use this new example to learn how to query the state of parfeval futures and cancel them. For more information, see Query and Cancel parfeval Futures (Parallel Computing Toolbox).

HTCondor integration: Plugin script for HTCondor now available as an Add-On

Integrate the third party scheduler HTCondor and MATLAB via the generic scheduler interface using an easy-to-configure plugin script. To download the plugin script, use the Add-On Explorer. Alternatively, you can download the plugin from the File Exchange. For more information, see Configure Using the Generic Scheduler Interface.

Parallel Workflows: Compare performance of different parallel environments

Learn how to choose when to use parallel pools, and when to use different parallel programming constructs. Use new examples to improve the performance of your parallel workflows:

- Compare Performance of Multithreading and ProcessPool (Parallel Computing Toolbox)
- Compare Performance of parfor, parfeval, and spmd (Parallel Computing Toolbox)

Query Underlying Data: Query the underlying data type of classes

You can now use the following functions to query the underlying data type of data stored in gpuArray objects, distributed arrays, dlarray objects, and more:

- underlyingType
- isUnderlyingType
- mustBeUnderlyingType

The class (MATLAB) function is useful to determine the class of a variable. However, some classes in MATLAB can contain underlying data that has a different type compared to what class returns. Example classes include gpuArray, dlarray, and distributed arrays. The underlyingType, isUnderlyingType, and mustBeUnderlyingType functions now provide a simple way to query the underlying data types of those classes.

For most classes, class(X) and underlyingType(X) return the same answer. However, for classes that can contain underlying data of a different type, class(X) returns the name of the class (such as gpuArray), and underlyingType(X) returns the underlying MATLAB data type that determines how the array X behaves (such as double).

Query Parallel Functionality: Query if support for Parallel Computing Toolbox functionality is available

You can now query if support for GPU and parallel pool functionality is available in your MATLAB installation using the following functions:

- canUseGPU (MATLAB)
- canUseParallelPool (MATLAB)

Use these functions to check supported functionality and avoid executing code that relies on specific hardware constraints.

Improved Scalability: Use MATLAB Job Scheduler clusters with up to 4000 workers

MATLAB Parallel Server with the MATLAB Job Scheduler now supports clusters up to 4000 workers. Support for large parallel pools remains at 1024 workers.

When you scale above 1000 workers, you must increase the heap memory available to the job manager. For more information, see Customize Startup Parameters.

Reference Architectures: Schedule jobs to run in AWS Batch

Use the MATLAB Parallel Server with AWS Batch reference architecture to build your own MATLAB Parallel Server solution for batch processing jobs using AWS Batch. AWS Batch dynamically provisions and manages Amazon EC2® instances based on the volume and resource requirements of the submitted jobs. AWS Batch supports both On-Demand and Spot Amazon EC2 instances.

To connect to the cluster from your MATLAB client, install the Parallel Computing Toolbox plugin for MATLAB Parallel Server with AWS Batch.

Docker Containers: Build a container image with a customized MATLAB installation

Use the reference Dockerfile to build a Docker container image that runs a custom MATLAB installation. You can choose the toolboxes you require in your custom installation. Containers are a scalable and reproducible method to deploy MATLAB in cloud and server environments. To access the Dockerfile and create a custom MATLAB container image, see Create a MATLAB Container Image.

labSend, labReceive, labBroadcast and gop Functions: Improved performance of data transfer between GPUs in a parallel pool

The labSend, labReceive, labBroadcast and gop functions show improved performance when transferring data in a parallel pool between MATLAB workers with GPUs. Data transfer between GPUs in a parallel pool now uses fast peer-to-peer communication, including NVLink, if available. For example, using the labSend and labReceive functions to transfer 72 MB data between two GPUs is approximate 5.1x faster than in the previous release.

```
% Set up parallel pool parpool('local',2);
% Clear all GPUs
spmd
    gpuDevice([]);
end
% Set sender/receiver worker index
senderIndex = 1:
receiverIndex = 2;
% Create data
     devices = gpuDevice();
    if labindex == senderIndex
    gpuData = ones(3000,3000,'gpuArray');
     end
    wait(devices):
% Measure time to transfer data 20 times
tic;
for j=1:20
    spmd
         if labindex == senderIndex
              labSend(gpuData,receiverIndex);
         elseif labindex == receiverIndex
   newGpuData = labReceive(senderIndex);
         end
         wait(devices):
     end
end
toc
% Delete parallel pool
delete(gcp('nocreate'))
```

The approximate execution times are:

R2020a: 9.50 seconds

R2020b: 1.85 seconds

The code was timed on a Windows 10, Intel Xeon E5-2623 v4 @ 2.60 GHz test system with four NVIDIA Titan V 12GB GPUs by running the above script.

Upgrade Parallel Computing Products Together

As with every new release, if you upgrade Parallel Computing Toolbox, you must upgrade MATLAB Parallel Server at the same time so that they interact properly with each other.

Compatibility Considerations

If you are using MATLAB Job Scheduler, the backward compatibility feature allows you to connect to multiple versions of MATLAB Parallel Server in your cluster. For more information, see Run Multiple MATLAB Parallel Server Versions (MATLAB Parallel Server).

Jobs created in one version of Parallel Computing Toolbox software do not run in a different version of MATLAB Parallel Server software, and might not be readable in different versions of the toolbox software. The job data stored in the folder identified by JobStorageLocation might not be compatible between different versions of MATLAB Parallel Server. You must specify a different JobStorageLocation for each parallel computing product, and each version on your cluster must have its own JobStorageLocation.

Functionality Being Removed or Changed

GPU acceleration is not available on macOS

Errors

Starting in R2020b, GPU acceleration, including gpuArray and CUDA functionality, is not supported on macOS platforms. Using gpuArray or CUDA functionality on macOS platforms results in an error.

For more information on platforms supported by MATLAB, see Platform Road Map for MATLAB and Simulink. For more information on GPU support for macOS, see Can I use MATLAB with an NVIDIA GPU on macOS 10.14 Mojave?

Forward compatibility for GPU devices is disabled by default

Behavior change

Starting in R2020b, forward compatibility for GPU devices is disabled by default. In previous releases, forward compatibility for GPU devices is enabled and cannot be disabled.

When forward compatibility is disabled, you cannot perform computations using a GPU device with an architecture that was released after the version of MATLAB you are using was built.

To reproduce the behaviour of earlier MATLAB versions, enable forward compatibility using the parallel.gpu.enableCUDAForwardCompatibility (Parallel Computing Toolbox) function or by setting the environment variable MW_CUDA_FORWARD_COMPATIBILITY to 1.

For more information, see Forward Compatibility for GPU Devices (Parallel Computing Toolbox).

classUnderlying and isaUnderlying are not recommended Still runs

classUnderlying (Parallel Computing Toolbox) and isaUnderlying (Parallel Computing Toolbox) are not recommended. Use underlyingType and isUnderlyingType instead.

R2020a

Version: 7.2

New Features

Compatibility Considerations

Parallel profiling: Learn tips and techniques to profile parallel code with new documentation

Use this new example to learn how to profile your parallel code and find out what functions parallel pool workers are spending most of the time on. For more information, see Profile Parallel Code.

Additional Properties Documentation: Learn how to customize the behavior of the sample plugin scripts

When you use the generic scheduler interface, you can modify the behavior of the plugin scripts by setting additional properties. For example, you can specify additional scheduler arguments for job submission. Follow these new instructions to learn more about the different additional properties that are available for the sample plugin scripts and how to set them: Customize Behavior of Sample Plugin Scripts.

Job Arrays: Submit job arrays to third-party schedulers with the generic scheduler interface

Now MATLAB leverages job arrays by default when you submit jobs to Slurm, LSF®, PBS Pro®, and Grid Engine using the generic scheduler interface. For more information on integrating a third-party scheduler with MATLAB and the generic scheduler interface, see Integrate MATLAB with Third-Party Schedulers and Configure Using the Generic Scheduler Interface.

GPU Functionality: Use new and enhanced gpuArray functions

- mldivide: Support for batch operations on rectangular matrices.
- tril: Support for batch operations.
- triu: Support for batch operations.
- qmr
- tfgmr

For more information, see Run MATLAB Functions on a GPU (Parallel Computing Toolbox).

GPU Functionality: Use new and enhanced gpuArray functions in Statistics and Machine Learning Toolbox

Over 140 functions in Statistics and Machine Learning Toolbox have support for gpuArrays. Functions with new gpuArray support include:

- corr
- random

For a full list of functions with new and enhanced GPU functionality, see *Release Notes for Statistics* and *Machine Learning Toolbox* (Statistics and Machine Learning Toolbox).

GPU Functionality: New gpuArray support for spectral functions

Several spectral functions in Audio Toolbox, Signal Processing Toolbox, and Wavelet Toolbox now support gpuArrays. The following functions now have gpuArray support:

- mfcc
- melSpectrogram
- czt
- spectrogram
- stft
- wvd, wvd
- cwt
- cwtfilterbank and wt object function

Distributed Arrays: Use new and enhanced distributed array functionality

- matches
- ilu
- del2
- inpolygon
- polyfit
- polyval
- renamevars
- write: support for new name-value pairs 'VariableEncoding' and 'Version' for writing Parquet files.
- subsasgn: support for indexed deletion.

For more information, see Run MATLAB Functions with Distributed Arrays (Parallel Computing Toolbox).

New Thread-Based Parallel Pool: Optimized for reduced memory usage, faster scheduling, and less data transfer, for a subset of MATLAB functions

You can now run parallel language features on a thread-based parallel pool. Thread-based environments are optimized for reduced memory usage, faster scheduling, and less data transfer. These environments are an alternative to the existing family of process-based environments.

Thread workers support a subset of the MATLAB functions available for process workers. If you are interested in a function that is not supported, let the MathWorks Technical Support team know.

The features that thread workers support are parpool, parfor, parfeval, tall, and parallel.pool.Constant. Features that are not supported include spmd, distributed, and parallel.pool.DataQueue.

In general, many core features of MATLAB are supported, including:

- · Language fundamentals
- Mathematics
- Core data types (double, single, logical, integer types, char, cell arrays, string, table, timetable, categorical, datetime, and duration)
- Control flow and logic (for example, if, for loops, and while loops)
- · Scripts, functions and classes
- Custom classes

In general, features that modify or access things outside of the thread worker are not supported, including:

- · Data import and export
- Graphics
- External languages

For more information, see Choose Between Thread-Based and Process-Based Environments (Parallel Computing Toolbox).

Upgrade Parallel Computing Products Together

As with every new release, if you upgrade Parallel Computing Toolbox you must upgrade MATLAB Parallel Server at the same time so that they interact properly with each other.

Compatibility Considerations

If you are using MATLAB Job Scheduler, the backward compatibility feature allows you to connect to multiple versions of MATLAB Parallel Server in your cluster. For more information, see Run Multiple MATLAB Parallel Server Versions.

Jobs created in one version of Parallel Computing Toolbox software do not run in a different version of MATLAB Parallel Server software, and might not be readable in different versions of the toolbox software. The job data stored in the folder identified by JobStorageLocation (formerly DataLocation) might not be compatible between different versions of MATLAB Parallel Server. Therefore, JobStorageLocation should not be shared by parallel computing products running different versions, and each version on your cluster should have its own JobStorageLocation.

Functionality Being Removed or Changed

pmode will be removed

Warns

In a future release, the pmode function will be removed. To execute commands interactively on multiple workers, use spmd instead.

pload and psave will be removed

Warns

In a future release, the psave and pload functions will be removed. To save and load data on the workers, in the form of Composite arrays or distributed arrays, use dsave and dload instead.

R2019b

Version: 7.1

New Features

GPU Functionality: Use new and enhanced gpuArray functions

- min, max: Support for the 'linear' option.
- diag, trace: Support for sparse gpuArrays.
- times, .*: Support for sparse gpuArrays.

For more information, see Run MATLAB Functions on a GPU (Parallel Computing Toolbox).

Support for NVIDIA CUDA 10.1: Update to CUDA Toolkit 10.1

The parallel computing products are now using CUDA Toolkit version 10.1. To compile CUDA code for CUDAKernel or CUDA MEX-files, you must use toolkit version 10.1. For more information, see GPU Support by Release (Parallel Computing Toolbox).

gpuArray: Load gpuArray data when no GPU is available

You can now load MAT files containing gpuArray data as in-memory arrays when a GPU is not available. gpuArrays loaded without a GPU are limited and you cannot use them for computations. To use a gpuArray loaded without a GPU, retrieve the contents with gather.

Distributed Arrays: Use new and enhanced distributed array functionality

- decomposition
- min, max: Support for the 'linear' option.
- mink, maxk, topkrows
- issorted, issortedrows
- distributed: Support for creating a distributed array from a Composite array.
- Support for accessing whole rows or columns in 2-D block-cyclic distributed arrays.

For more information, see Run MATLAB Functions with Distributed Arrays (Parallel Computing Toolbox).

Distributed Arrays Examples: Explore multigrid preconditioned iterative solvers with new documentation

Use this new example to learn how to use a multigrid preconditioned iterative solver with distributed arrays. For more information, see Solve Differential Equation Using Multigrid Preconditioner on Distributed Discretization (Parallel Computing Toolbox).

Parallel Deep Learning Workflows: Explore deep learning with custom parallel training loops

Use this new example to learn how to set up custom training loops to train deep learning networks in parallel. For more information, see Train Network in Parallel with Custom Training Loop (Parallel Computing Toolbox).

Batch Jobs Examples: Explore batch workflows with new documentation

Use these new examples to learn about batch jobs and options:

- Run Script as Batch Job (Parallel Computing Toolbox)
- Run Batch Job and Access Files from Workers (Parallel Computing Toolbox)

parfeval Examples: Explore parfeval workflows with new documentation

Use these new examples to learn more about parfeval workflows:

- Plot During Parameter Sweep with parfeval (Parallel Computing Toolbox)
- Perform Webcam Image Acquisition in Parallel with Postprocessing (Parallel Computing Toolbox)
- Perform Image Acquisition and Parallel Image Processing (Parallel Computing Toolbox)

Image Acquisition Examples: Perform parallel acquisition and processing of images

Use these new examples to learn how to implement these parallel workflows:

- Perform Webcam Image Acquisition in Parallel with Postprocessing (Parallel Computing Toolbox)
- Perform Image Acquisition and Parallel Image Processing (Parallel Computing Toolbox)

Improved Scalability: Use MATLAB Job Scheduler clusters with up to 2000 workers

MATLAB Parallel Server with the MATLAB Job Scheduler now supports clusters up to 2000 workers. Support for large parallel pools remains at 1024 workers.

Third-Party Schedulers: Check the scheduler ID of a MATLAB task

If you submit a task to a third-party scheduler, then you can check its ID on the scheduler by looking at the SchedulerID property of the task object. For more information, see CJS Tasks (Parallel Computing Toolbox). To get the scheduler IDs of all tasks in a job, use getTaskSchedulerIDs.

Common Job Schedulers: Improved performance of vectorized task creation

Local and third-party schedulers benefit from improved task creation rates and submission times when you use createTask for vectorized task creation or parfor without a parallel pool.

For example, the following functions show about a 75x speed-up for task creation and a 6x speed-up for job submission.

```
function taskCreationTimingTest
    c = parcluster('local');
```

```
i = createJob(c);
    numTasks = 10000;
    allTaskArgs = cell(1,numTasks);
    for ii = 1:numTasks
        ithTaskArgs = {ii}; % Arguments to the ith task
        allTaskArgs{ii} = ithTaskArgs;
    end
    tic;
    createTask(j,@rand,1,allTaskArgs);
    toc
end
function jobSubmissionTimingTest
    c = parcluster('local');
    i = createJob(c);
    numTasks = 10000;
   allTaskArgs = cell(1,numTasks);
    for ii = 1:numTasks
        ithTaskArgs = {ii}; % Arguments to the ith task
        allTaskArgs{ii} = ithTaskArgs;
    createTask(j,@rand,1,allTaskArgs);
    tic;
    submit(j);
    toc
end
```

The approximate times are:

- R2019a: 296 s for task creation and 150 s for job submission.
- R2019b: 3.9 s for task creation and 24 s for job submission.

The code was timed on a Windows 10, Intel Xeon W-2133 CPU @ 3.60 GHz test system using the timeit function:

```
timeit(@taskCreationTimingTest)
timeit(@jobSubmissionTimingTest)
```

Personalized Clusters: Set Cloud Center clusters for personal use

Now Cloud Center clusters can have two levels of authorization: for personal use or sharable.

Upgrade Parallel Computing Products Together

As with every new release, if you upgrade Parallel Computing Toolbox you must upgrade MATLAB Parallel Server at the same time so that they interact properly with each other.

Compatibility Considerations

If you are using MATLAB Job Scheduler, the backward compatibility feature allows you to connect to multiple versions of MATLAB Parallel Server in your cluster. For more information, see Run Multiple MATLAB Parallel Server Versions.

Jobs created in one version of Parallel Computing Toolbox software do not run in a different version of MATLAB Parallel Server software, and might not be readable in different versions of the toolbox

software. The job data stored in the folder identified by JobStorageLocation (formerly DataLocation) might not be compatible between different versions of MATLAB Parallel Server. Therefore, JobStorageLocation should not be shared by parallel computing products running different versions, and each version on your cluster should have its own JobStorageLocation.

Functionality Being Removed or Changed

Toolbox folder renamed from distcomp to parallel

Behavior change

In R2019b, the Parallel Computing Toolbox folder has been renamed. The new name for the toolbox folder is parallel. If you need to reference the location of the toolbox, update your references to use toolbox/parallel instead of toolbox/distcomp.

pmode will be removed

Still runs

In a future release, the pmode function will be removed. To execute commands interactively on multiple workers, use spmd instead.

pload and psave will be removed

Still runs

In a future release, the psave and pload functions will be removed. To save and load data on the workers, in the form of Composite arrays or distributed arrays, use dsave and dload instead.

R2019a

Version: 7.0

New Features

Updated license model: Scale beyond 200 workers without checking out more licenses

If you have 200 workers checked out and are not using on-demand licensing, you can continue to scale further without checking out more workers. To learn more, see License Model.

Renamed Product: MATLAB Distributed Computing Server renamed to MATLAB Parallel Server

MATLAB Distributed Computing Server now has the name MATLAB Parallel Server.

parfor-Loops: Use parfor without a parallel pool

parfor can now run without a parallel pool of workers. If you choose this approach, parfor uses the available cluster resources as needed. You can also control options, such as the worker load, with parforOptions.

Automatic Cluster Resizing: Resize Cloud Center clusters on Amazon based on usage

You can now create cloud clusters that resize automatically based on usage. These clusters grow or shrink to allocate the optimal number of workers for your submitted tasks. For more information, see the MathWorks Cloud Center documentation.

Benchmarks: Use new examples to evaluate the performance of your cluster

Use the computational tests in these benchmarks to measure the performance of your compute cluster.

- Benchmark Cluster Workers (Parallel Computing Toolbox)
- Benchmark Your Cluster with the HPC Challenge (Parallel Computing Toolbox)

parpool Resiliency: Parallel pools are now resilient to dropped network connections

Parallel pools are now robust to network failures, and can recover from connectivity issues. For example, if your network connection drops, MATLAB attempts to reconnect to the workers in the pool. There is no loss of data if the disconnection happens during data transfer.

Parallel Deep Learning Workflows: Explore deep learning with multiple-GPUs

Use these new examples to learn how to use multiple GPUs to train a single or several deep learning networks in parallel.

• Train Network Using Automatic Multi-GPU Support (Parallel Computing Toolbox)

• Run Multiple Deep Learning Experiments (Parallel Computing Toolbox)

Custom Datastore: Read from Hadoop based databases using the custom datastore framework

Author a custom database datastore and access data from a database of your choice using these extensions:

- matlab.io.datastore.HadoopLocationBased Use this mixin to specify the location of your data in Hadoop®.
- matlab.io.datastore.HadoopInput Use this helper class when a Hadoop java class is available for a database library.

Perform big data analysis on your custom datastore with tall arrays or mapreduce. These custom datastores can leverage the location of the data to make computations more efficient.

For more information on the custom datastore framework, see Develop Custom Datastore (MATLAB).

GPU Functionality: Use new and enhanced gpuArray functions

- sinpi
- cospi
- normalize
- validateattributes

For more information, see Run MATLAB Functions on a GPU (Parallel Computing Toolbox).

Support for NVIDIA CUDA 10.0: Update to CUDA Toolkit 10.0

The parallel computing products are now using CUDA Toolkit version 10.0. To compile CUDA code for CUDAKernel or CUDA MEX-files, you must use toolkit version 10.0. For more information, see GPU Support by Release (Parallel Computing Toolbox).

Distributed Arrays: Use new and enhanced distributed array functionality

- sinpi
- cospi
- normalize
- validateattributes
- write: Support for writing to Parguet files
- eig: Support for non-symmetric matrices
- sprandsym
- Distributed iterative solvers, such as pcg or gmres, now allow arbitrary matrices as preconditioners

For more information, see Run MATLAB Functions with Distributed Arrays (Parallel Computing Toolbox).

Reference Architectures: Support for network license manager

Use reference architectures to serve a network license manager for your MATLAB Parallel Server cloud clusters. Use preconfigured server settings for Amazon AWS and Microsoft® Azure.

- MATLAB Parallel Server on Amazon Web Services.
- License Manager for MATLAB on Amazon Web Services.
- MATLAB Parallel Server on Microsoft Azure.
- License Manager for MATLAB on Microsoft Azure.

Support for MPICH: Update to MPICH Version 3 on Linux for thirdparty schedulers

The parallel computing products are now shipping MPICH version 3.2.1 on Linux for third-party schedulers. Communicating jobs for third-party schedulers now use the Hydra process manager.

Compatibility Considerations

If you use your own MPI builds, you might need to create new builds compatible with this latest version. For instructions, see Use Different MPI Builds on UNIX Systems.

Upgrade Parallel Computing Products Together

As with every new release, if you upgrade Parallel Computing Toolbox you must upgrade MATLAB Parallel Server at the same time so that they interact properly with each other.

Compatibility Considerations

If you are using MATLAB Job Scheduler, the backward compatibility feature allows you to connect to multiple versions of MATLAB Parallel Server in your cluster. For more information, see Run Multiple MATLAB Parallel Server Versions.

Jobs created in one version of Parallel Computing Toolbox software do not run in a different version of MATLAB Parallel Server software, and might not be readable in different versions of the toolbox software. The job data stored in the folder identified by JobStorageLocation (formerly DataLocation) might not be compatible between different versions of MATLAB Parallel Server. Therefore, JobStorageLocation should not be shared by parallel computing products running different versions, and each version on your cluster should have its own JobStorageLocation.

Functionality Being Removed or Changed

Default random number generator changed for parallel contexts *Behavior change*

Starting in R2019a, the default random number generator for parallel computations is changed to Threefry. This change applies to calculations on parallel workers, GPU arrays, distributed arrays,

and tall arrays. This generator offers performance enhancements for parallel calculations over the previous default. In releases up to and including R2018b, the default random number generator for parallel computations is CombRecursive.

With a different default generator, MATLAB generates different random number sequences by default in the context of parallel computations. However, the statistics of these calculations remain unaffected. Therefore, you might want to update any code that relies on the specific random numbers being generated, but most calculations on the random numbers are unaffected.

To set the generator to the settings used by default in R2018b and earlier on parallel workers, GPU arrays, and tall arrays, use the following commands.

Calculations on parallel workers and distributed arrays:

```
spmd
    stream = RandStream.create("CombRecursive", "NumStreams", 2^32,...
        "StreamIndices", 2*labindex);
    RandStream.setGlobalStream(stream);
end

Calculations on GPU arrays:
gpurng(0,"CombRecursive")

Calculations on tall arrays:
tallrng(0,"CombRecursive")
```

The remotecopy command will no longer support the rcp protocol in a future release Warns

The rcp protocol is insecure. The remotecopy function will no longer support rcp as an option to the -protocol flag in a future release, and warns starting in R2018b. Instead, specify any other of the supported protocols, such as scp. For a complete list, see remotecopy.

The remotemjs command will no longer support the rsh protocol in a future release Warns

The rsh protocol is insecure. The remotemjs function will no longer support rsh as an option to the -protocol flag in a future release, and warns starting in R2018b. Instead, specify any other of the supported protocols, such as ssh. For a complete list, see remotemjs.

The mdce command will be removed in a future release Warns

mdce will be removed in a future release. Use mis instead.

The defaults file is now mjs def. To specify an alternative file, you can use the -mjsdef flag.

For more information, see mis.

The remotemdce command will be removed in a future release *Warns*

remotemdce will be removed in a future release. Use remotemjs instead. For more information, see remotemjs.

Toolbox folder renamed from distcomp to parallel

Behavior change in future release

In R2019b, the Parallel Computing Toolbox toolbox folder will be renamed. The new name for the toolbox folder is parallel. If you need to reference the location of the toolbox, update your references to use toolbox/parallel instead of toolbox/distcomp.

R2018b

Version: 6.13

New Features

GPU Support: View details of GPU support on over 600 function pages, and browse GPU support for functions by toolbox

Consult additional GPU usage information on function pages of GPU-enabled functions in MATLAB and other toolboxes. You can find this information in the Extended Capabilities section at the end of the function page. Also, browse GPU function lists filtered by product. For more information, see Run MATLAB Functions on a GPU (Parallel Computing Toolbox).

GPU Functionality: Use new and enhanced gpuArray functions, such as spline interpolation and vecnorm

- interp1: support for 'spline' interpolation method
- vecnorm
- norm for sparse gpuArrays: support for all vector norms; and 1, Inf, and frobenius matrix norms
- Reduction functions (min, max, sum,...): support for reducing multiple dimensions simultaneously
- nthroot: support for logical inputs
- sign: support for logical inputs

For more information, see Run MATLAB Functions on a GPU (Parallel Computing Toolbox).

Support for NVIDIA CUDA 9.1: Update to CUDA Toolkit 9.1

The parallel computing products are now using CUDA Toolkit version 9.1. To compile CUDA code for CUDAKernel or CUDA MEX-files, you must use toolkit version 9.1. For more information, see GPU Support by Release (Parallel Computing Toolbox).

GPU Optimizations: Enhanced support and optimizations for GPU

- gpuDevice and gpuDeviceCount now run faster if you do not have a GPU, because they do not load the GPU libraries. You can use these functions to check if a setup supports GPUs or not.
- GPU memory management is now better optimized. Benefit from improved performance in memory intensive GPU applications, such as deep learning.

GPU Examples: Explore GPU workflows on single and multiple GPUs

New and updated GPU documentation now includes:

- Use MATLAB Functions with a GPU (Parallel Computing Toolbox)
- Identify and Select a GPU Device (Parallel Computing Toolbox)
- Sharpen an Image Using the GPU (Parallel Computing Toolbox)
- Run MATLAB Functions on Multiple GPUs (Parallel Computing Toolbox)
- Use Multiple GPUs in a Parallel Pool (Parallel Computing Toolbox)
- qpuArray
- Establish Arrays on a GPU (Parallel Computing Toolbox)
- GPU Support by Release (Parallel Computing Toolbox)

Distributed Arrays: Use new and enhanced distributed array functionality, including support for vecnorm and writing to Amazon S3 and Azure

- vecnorm
- write: support for writing to XLS, CSV, and TXT formats
- write: extended support for writing SEQ and MAT formats to all supported file systems
- write: support for writing to S3 and WASBS file systems
- Reduction functions (min, max, sum,...): support for reducing multiple dimensions simultaneously
- nthroot: support for logical inputs
- sign: support for logical inputs
- Distributed timetable support
- mldivide: more robust sparse distributed system solve

For more information, see Run MATLAB Functions with Distributed Arrays (Parallel Computing Toolbox).

Distributed Support: View details of distributed array support on over 400 function pages, and browse distributed support for functions by toolbox

Consult additional distributed array usage information on function pages of distributed-enabled functions in MATLAB and other toolboxes. You can find this information in the Extended Capabilities section at the end of the function page. Also, browse distributed array function lists filtered by product. For more information, see Run MATLAB Functions with Distributed Arrays (Parallel Computing Toolbox).

Parallel Extended Capabilities: View details of automatic parallel support on function pages, and browse support for functions by toolbox

Many functions in MATLAB, Simulink, and toolboxes help you take advantage of parallel computing resources without requiring any extra coding. You can enable this support by simply setting a flag or preference.

Now you can consult automatic parallel usage information on function pages and browse supported function lists filtered by product.

Similarly, you can view lists of functions with extended capabilities for GPU arrays and Distributed arrays enabled by Parallel Computing Toolbox.

For details, see

- Run MATLAB Functions with Automatic Parallel Support (Parallel Computing Toolbox)
- Run MATLAB Functions on a GPU (Parallel Computing Toolbox)
- Run MATLAB Functions with Distributed Arrays (Parallel Computing Toolbox)

Write Cloud Data: Write distributed arrays and tall arrays to Amazon S3 and Windows Azure storage

If you have access to cloud storage in Amazon $S3^{\text{m}}$ or Windows Azure Blob Storage, you can upload your distributed and tall data using the write function. When you write data from a cluster in the same cloud service, the data transfer is more efficient. To learn more about your options for accessing cloud storage, see Work with Remote Data (MATLAB).

Big Data in the Cloud: Explore MATLAB capabilities for big data

A new example shows how to access a publicly available dataset in the cloud and work with it using datastores, tall arrays and Parallel Computing Toolbox. This example shows you step-by-step how to use MATLAB to analyse huge datasets. See Process Big Data in the Cloud (Parallel Computing Toolbox).

Streamlined Cloud Cluster Setup: Create new Cloud Center clusters directly from the MATLAB desktop

You can now create a MATLAB Distributed Computing Server enabled cluster on Amazon EC2 directly from MATLAB. On the **Home** tab, in the **Environment** area, select **Parallel > Create and Manage Clusters**, and choose **Create Cloud Cluster** to set up a new cloud cluster. For more information, see Create Cloud Cluster (Parallel Computing Toolbox).

Improved Scalability: Use up to 1024 workers per parallel pool

MATLAB Distributed Computing Server can now support clusters with up to 1024 workers. Use large parallel pools of workers for big scaling problems.

Reference Architectures: Create customized MATLAB Distributed Computing Server clusters in the cloud using Amazon Web Services (AWS) or Microsoft Azure

Use reference architectures to build your own MATLAB Distributed Computing Server cluster solution for the cloud. Use preconfigured cluster settings for Amazon AWS and Microsoft Azure. For more information on launching and configuring clusters, see

- MATLAB Parallel Server on Amazon Web Services.
- MATLAB Parallel Server on Microsoft Azure.

For more information, see MATLAB in the Cloud (https://www.mathworks.com/cloud.html).

Cluster Workflows: Learn how to scale up from desktop to cluster

Use this new example to learn how to prototype interactively your parallel code on your local machine and scale up to a cluster. For more information, see Scale up from Desktop to Cluster (Parallel Computing Toolbox).

Streamlined Installation Documentation: Simplified workflows for integrating MATLAB Distributed Computing Server in your cluster

Follow simplified workflows to integrate your third-party scheduler or MATLAB Job Scheduler with MATLAB Distributed Computing Server in your cluster. For details, see Getting Started with MATLAB Distributed Computing Server.

Generic Scheduler Documentation: Set up Schedulers Using Improved Instructions

Follow improved instructions to integrate a third-party scheduler using the generic scheduler interface. For more information, see Configure Using the Generic Scheduler Interface.

Parallel Deep Learning Workflows: Explore deep learning with multiple GPUs locally or in the cloud

Use examples to explore options for scaling up deep learning training. You can use multiple GPUs locally or in the cloud without changing your code. Use parallel computing to train multiple networks locally or on cloud clusters, and use datastores to access cloud data. New examples include:

- Train Network in the Cloud Using Built-in Parallel Support (Parallel Computing Toolbox)
- Use parfor to Train Multiple Deep Learning Networks (Parallel Computing Toolbox)
- Use parfeval to Train Multiple Deep Learning Networks (Parallel Computing Toolbox)
- Upload Deep Learning Data to the Cloud (Parallel Computing Toolbox)
- Send Deep Learning Batch Job To Cluster (Parallel Computing Toolbox)

To learn about options, see Scale Up Deep Learning in Parallel and in the Cloud (Deep Learning Toolbox).

Upgrade Parallel Computing Products Together

As with every new release, if you upgrade Parallel Computing Toolbox you must upgrade MATLAB Distributed Computing Server at the same time so that they interact properly with each other.

Compatibility Considerations

If you are using MATLAB Job Scheduler, the backward compatibility feature allows you to connect to multiple versions of MATLAB Distributed Computing Server in your cluster. For more information, see Run Multiple MATLAB Distributed Computing Server Versions.

Jobs created in one version of Parallel Computing Toolbox software do not run in a different version of MATLAB Distributed Computing Server software, and might not be readable in different versions of the toolbox software. The job data stored in the folder identified by JobStorageLocation (formerly DataLocation) might not be compatible between different versions of MATLAB Distributed Computing Server. Therefore, JobStorageLocation should not be shared by parallel computing products running different versions, and each version on your cluster should have its own JobStorageLocation.

R2018a

Version: 6.12

New Features

Parfeval callbacks: New afterAll and afterEach methods for parallel futures

You can use the new methods afterAll and afterEach to specify functions to automatically execute when parallel futures complete. You create parallel futures using the parfeval function. A useful application for afterEach and afterAll is to update user interfaces such as plots and apps during parallel computations using parfeval. For an example, see Update a User Interface Asynchronously Using afterEach and afterAll.

Slurm support: Slurm is a fully supported scheduler

Slurm is now a fully supported scheduler. You can create Slurm profiles directly in the Profile Manager without needing templates and scripts from File Exchange. For more information, see Configure for Slurm, PBS Pro, Platform LSF, TORQUE.

Support for NVIDIA Volta: Update to CUDA 9, support for Volta class GPUs

The parallel computing products are now using CUDA Toolkit version 9.0. To compile CUDA code for CUDAKernel or CUDA MEX-files, you must use toolkit version 9.0.

This update improves support for Volta class GPUs.

Improved file mirroring: Performance of file mirroring for generic scheduler integration

File mirroring for the generic scheduler integration has been improved through the use of zip archives.

Parfor performance improvements: More efficient broadcast variables in parfor for non-local clusters

If you use broadcast variables in parfor loops with non-local clusters, now performance is improved. Broadcast variables in parfor loops are sent only once to the cluster, instead of once per worker.

GPU Array Support: Use enhanced gpuArray functions

You can now use the following enhanced gpuArray functions:

- rescale
- new syntaxes of sort and sortrow

For more information, see Run Built-In Functions on a GPU.

Distributed Array Support: Use enhanced distributed array functions

You can now use the following enhanced distributed array functions:

- rescale
- new syntaxes of sort and sortrow
- bicgstabl

For more information, see Using MATLAB Functions on Distributed Arrays.

Parameter Sweep Example: Use a DataQueue to monitor results during computations on a parallel pool

The new example *Plot during Parameter Sweep with parfor* shows how to perform a parameter sweep in parallel and plot progress during parallel computations. For details, see Plot during Parameter Sweep with parfor.

Discontinued Support for GPU Devices of Compute Capability less than 3.0

In R2018a, support for GPU devices of compute capability less than 3.0 is removed. A minimum compute capability of 3.0 is required for GPU computing in MATLAB. For more information on using GPUs, see GPU Computing in MATLAB.

Compatibility Considerations

In R2018a, any use of **gpuDevice** to select a GPU with compute capability less than 3.0 generates an error. Support is removed for devices with compute capability less than 3.0.

Backwards Compatibility for MATLAB Job Scheduler: Upgrade MATLAB Job Scheduler clusters, and continue to use previous releases of Parallel Computing Toolbox

You can upgrade your MATLAB Job Scheduler clusters to the latest release and still connect to it using previous releases (back to R2016a) of Parallel Computing Toolbox on your MATLAB desktop client. For the supported older releases, you must maintain an installation of the same release of MATLAB Distributed Computing Server for each release of MATLAB you want to use. The latest MATLAB Job Scheduler routes your jobs accordingly. You can configure MATLAB Job Scheduler with the location of these installations in the mdce_def file. For more information, see Install Products and Choose Cluster Configuration.

Upgrade Parallel Computing Products Together

As with every new release, if you upgrade Parallel Computing Toolbox you must upgrade MATLAB Distributed Computing Server at the same time so that they interact properly with each other. Note that you do not need to do this if you are taking advantage of MATLAB Job Scheduler's backwards compatibility.

The R2018a version of Parallel Computing Toolbox software is accompanied by a corresponding new version of MATLAB Distributed Computing Server software.

Compatibility Considerations

As with every new release, if you are using both parallel computing products, you must upgrade Parallel Computing Toolbox and MATLAB Distributed Computing Server together. These products must be the same version to interact properly with each other. This requirement applies only if you are not taking advantage of MATLAB Job Scheduler backwards compatibility.

Jobs created in one version of Parallel Computing Toolbox software do not run in a different version of MATLAB Distributed Computing Server software, and might not be readable in different versions of the toolbox software. The job data stored in the folder identified by JobStorageLocation (formerly DataLocation) might not be compatible between different versions of MATLAB Distributed Computing Server. Therefore, JobStorageLocation should not be shared by parallel computing products running different versions, and each version on your cluster should have its own JobStorageLocation.

Functionality Being Removed or Changed

Functionality	Result	Use Instead	Compatibility Considerations
Support for GPU devices of compute capability less than 3.0 is removed.	Errors	Use GPUs with a minimum compute capability of 3.0.	Support is removed for GPU devices of compute capability 2.x. Use GPUs with a minimum compute capability 3.0. Use the function gpuDevice to query the compute capabilities of your GPUs. For more information, see GPU Computing in MATLAB.
parallel.gpu.rng is renamed to gpurng	Still runs	gpurng	Replace all instances of parallel.gpu.rng with gpurng

R2017b

Version: 6.11

New Features

Bug Fixes

Improved Parallel Language Performance: Execute parallel language constructs with reduced overhead

All parallel language constructs, including parfor, run with reduced overhead, particularly constructs with short duration. The scheduling of parfor-loop intervals has been changed to utilize the cluster hardware more efficiently.

Tall Array Support: Use tall arrays with Windows client access to Linux Spark clusters

You can use tall arrays on Spark enabled Hadoop clusters supporting all architectures for the client, while supporting Linux and Mac architectures for the cluster. This includes cross-platform support.

For more detail on Hadoop clusters, see Configure a Hadoop Cluster.

Improved Parallel Pool Robustness: Run pools without Message Passing Interface (MPI) by default, making pools resilient to workers crashing

You can now run parallel pools without MPI by default, improving the resilience of your parallel pool to workers crashing.

Improved MATLAB Integration with Third-Party Schedulers: Use the Generic Profile Wizard for easier installation and setup of MATLAB Distributed Computing Server

You can now use the new workflow to automatically create generic profiles using the support packages for third-party schedulers. For more details, see Distribute a Generic Cluster Profile and Integration Scripts.

Cloud Storage: Work with data in Windows Azure Blob Storage

You can now use Windows Azure[®] Blob Storage to access data in the cloud using datastore. For more information on this topic, see Read Remote Data.

Copy Client Environment to Workers on Any Cluster: Specify which environment variables on your client machine your workers should automatically inherit

You can now mirror any of your client environment variables to the workers on a cluster. For example, you can include any environment variables required by third-party software. Or you can include cloud service provider credentials in your local environment to give worker processes access to your data stored in the cloud.

You can now set EnvironmentVariables using parpool, batch, createJob, or in the Cluster Profile Manager. EnvironmentVariables is also a common property of the parallel. Job class.

Client Path Sharing: Add user-added-entries on the client's path to the workers' paths

You can now automatically add user-added-entries on the client's path to the workers' paths for batch and independent jobs. This functionality can be enabled or disabled by specifying the AutoAddClientPath option to the parpool, batch, and createJob commands. AutoAddClientPath is also a common property of the parallel.Job class.

GPU Array Support: Use enhanced gpuArray functions

You can now use the following enhanced gpuArray functions:

- bounds for computing max and min simultaneously
- Sparse iterative solvers cgs and lsgr
- spdiags

For more information on this topic, see Run Built-In Functions on a GPU (Parallel Computing Toolbox).

Distributed Array Support: Use enhanced distributed array functions

You can now use the following enhanced distributed array functions:

- Sparse iterative solvers minres, symmlq and bicgstab
- bounds for computing max and min simultaneously
- eigs
- spdiags

For more information, see Using MATLAB Functions on Distributed Arrays (Parallel Computing Toolbox).

Enhanced Support for Microsoft Windows HPC Pack

The parallel computing products now support Microsoft Windows HPC Pack 2016. For details, see Configure for HPC Pack.

Discontinued Support for GPU Devices of Compute Capability less than 3.0

In a future release, support for GPU devices of compute capability less than 3.0 will be removed. At that time, a minimum compute capability of 3.0 will be required.

Compatibility Considerations

In R2017b, any use of gpuDevice to select a GPU with compute capability less than 3.0, generates a warning. The device is still supported in this release, but in a future release support will be completely removed for devices with compute capability less than 3.0.

Backwards Compatibility for MATLAB Job Scheduler: Upgrade MATLAB Job Scheduler clusters, and continue to use previous releases of Parallel Computing Toolbox

You can upgrade your MATLAB Job Scheduler clusters to R2017b and still connect to it using the R2017a, R2016b, and R2016a releases of Parallel Computing Toolbox on your MATLAB desktop client. For releases prior to R2016b, you must maintain an installation of the same release of MATLAB Distributed Computing Server for each release of MATLAB you want to use. The latest MATLAB Job Scheduler routes your jobs accordingly. You can configure MATLAB Job Scheduler with the location of these installations in the mdce_def file. For more information, see Install Products and Choose Cluster Configuration.

Upgrade Parallel Computing Products Together

The R2017b version of Parallel Computing Toolbox software is accompanied by a corresponding new version of MATLAB Distributed Computing Server software.

Compatibility Considerations

As with every new release, if you are using both parallel computing products, you must upgrade Parallel Computing Toolbox and MATLAB Distributed Computing Server together. These products must be the same version to interact properly with each other. This requirement applies only if you are not taking advantage of MATLAB Job Scheduler backwards compatibility.

Jobs created in one version of Parallel Computing Toolbox software do not run in a different version of MATLAB Distributed Computing Server software, and might not be readable in different versions of the toolbox software. The job data stored in the folder identified by JobStorageLocation (formerly DataLocation) might not be compatible between different versions of MATLAB Distributed Computing Server. Therefore, JobStorageLocation should not be shared by parallel computing products running different versions, and each version on your cluster should have its own JobStorageLocation.

Functionality Being Removed or Changed

Functionality	Result		Compatibility Considerations
Support for running MATLAB MapReduce on Hadoop 1.x clusters has been removed.		Hadoop 2.x or higher installed to run	Migrate MATLAB MapReduce code that runs on Hadoop 1.x to Hadoop 2.x.

For more information, see Configure a Hadoop Cluster.

R2017a

Version: 6.10

New Features

Bug Fixes

Backwards Compatibility for MATLAB Job Scheduler: Upgrade MATLAB Job Scheduler clusters, and continue to use previous releases of Parallel Computing Toolbox

You can upgrade your MATLAB Job Scheduler clusters to R2017a and still use the R2016b and R2016a releases of Parallel Computing Toolbox on your MATLAB desktop client to connect to it. This situation only applies to the R2016a release onward. You must maintain an installation of the same release of MATLAB Distributed Computing Server for each release of MATLAB you want to use. The latest MATLAB Job Scheduler routes your jobs accordingly. You can configure MATLAB Job Scheduler with the location of these installations in the mdce_def file. For more information, see Install Products and Choose Cluster Configuration.

Simplified Integration for Third-Party Cluster Schedulers: Updates to generic scheduler integration allow folder-based configuration and eliminate the need to specify function handles

You no longer need to specify function handles in your generic scheduler profile. Instead, specify only one folder name and some AdditionalProperties, which are similar to name-value pairs. For more information, see Configure for a Generic Scheduler.

Ability to Re-create Parallel Jobs: Easily rerun all failed or cancelled tasks for a job

If your job failed or was cancelled, you can now use the recreate method to return a new job object. Call submit on the new job object to easily rerun all failed or cancelled tasks.

More Responsive Job Monitor: Automatic updates for new, submitted, or deleted jobs or tasks

You can now use the Job Monitor to get an up-to-date view of your cluster. Verify your changes without manually querying the cluster or forcing an update to the Job Monitor user interface. Examine your latest changes and execute quickly without interrupting your workflow. For more details, see Job Monitor (Parallel Computing Toolbox).

Access to Intermediate Results and Updates in Parallel Computations: Poll for messages or data from different workers during parallel workflows

You can now transfer intermediate results when you carry out parfor, spmd, or parfeval calculations. Use the send and poll methods together to send and poll for messages or data from different workers using a DataQueue.

Distributed Array Support: Use enhanced distributed array functions

You can now use the following enhanced distributed array functions:

• issymmetric, ishermitian

- qmr
- svds
- tfgmr
- write for storing a snapshot of a distributed array in a format suitable for datastore
- mldivide, providing improved performance for triangular and diagonal systems

For more information, see Using MATLAB Functions on Distributed Arrays (Parallel Computing Toolbox).

Support for Spark 2.x enabled Hadoop clusters

Tall array integration on a Spark enabled Hadoop cluster, running MATLAB Distributed Computing Server, now supports both versions 1.x and 2.x.

Discontinued Support for GPU Devices of Compute Capability less than 3.0

In a future release, support for GPU devices of compute capability less than 3.0 will be removed. At that time, a minimum compute capability of 3.0 will be required.

Compatibility Considerations

In R2017a, any use of <code>gpuDevice</code> to select a GPU with compute capability less than 3, generates a warning. The device is still supported in this release, but in a future release support will be completely removed for devices with compute capability less than 3.

Upgrade parallel computing products together

This version of MATLAB Distributed Computing Server software is accompanied by a corresponding new version of Parallel Computing Toolbox software.

Compatibility Considerations

As with every new release, if you are using both parallel computing products, you must upgrade Parallel Computing Toolbox and MATLAB Distributed Computing Server together. These products must be the same version to interact properly with each other.

Jobs created in one version of Parallel Computing Toolbox software do not run in a different version of MATLAB Distributed Computing Server software, and might not be readable in different versions of the toolbox software. The job data stored in the folder identified by JobStorageLocation (formerly DataLocation) might not be compatible between different versions of MATLAB Distributed Computing Server. Therefore, JobStorageLocation should not be shared by parallel computing products running different versions, and each version on your cluster should have its own JobStorageLocation.

Properties of generic cluster objects have changed

The properties of new generic cluster objects have changed.

Compatibility Considerations

From R2017a onwards, you can no longer use the following old properties:

- CancelJobFcn
- CancelTaskFcn
- CommunicatingSubmitFcn
- DeleteJobFcn
- DeleteTaskFcn
- GetJobStateFcn
- IndependentSubmitFcn

Instead, set the new IntegrationScriptsLocation property to the folder containing your cluster's integration scripts. You must now use integration scripts with the default function name and signature. Specify any additional input arguments to these scripts using the new AdditionalProperties property. For more details, see

- · Configure for a Generic Scheduler
- · Program Communicating Jobs for a Generic Scheduler
- · Program Independent Jobs for a Generic Scheduler

Functionality Being Removed or Changed

Functionality	Result	Use Instead	Compatibility Considerations
Support for running MATLAB MapReduce on Hadoop 1.x clusters will be removed in a future release.		Use clusters that have Hadoop 2.x or higher installed to run MATLAB MapReduce.	Migrate MATLAB MapReduce code that runs on Hadoop 1.x to Hadoop 2.x.

For more information, see Configure a Hadoop Cluster.

R2016b

Version: 6.9

New Features

Bug Fixes

Backwards Compatibility: Upgrade MATLAB Job Scheduler clusters, and continue to use the previous release of Parallel Computing Toolbox

You can upgrade your MATLAB Job Scheduler clusters to R2016b and still use the R2016a release of Parallel Computing Toolbox on your MATLAB desktop client to connect to it. This situation only applies to the R2016a release onward. You must maintain an installation of the same release of MATLAB Distributed Computing Server for each release of MATLAB you want to use. The latest MATLAB Job Scheduler will route your jobs accordingly. You can configure MATLAB Job Scheduler with the location of these installations in the mdce_def file. For more information, see Install Products and Choose Cluster Configuration.

Cluster Profile Validation: Choose which validation stages run and the number of MATLAB workers to use

In previous releases, validating your cluster profile ran all validation stages and used a fixed number of workers determined from your profile. You can now choose to run a subset of the validation stages and specify the number of workers to use when validating your profile. For more information on the detailed validation steps in your cluster profile, see Validate Cluster Profiles.

Parallel Support for Tall Arrays: Process big data with tall arrays in parallel on your desktop, MATLAB Distributed Computing Server, and Spark clusters

Speed up your tall array workflows with Parallel Computing Toolbox. You can use Parallel Computing Toolbox to evaluate tall array expressions in parallel using a parallel pool on your desktop. You can also use Parallel Computing Toolbox to scale up tall-array processing by connecting to a parallel pool running on a MATLAB Distributed Computing Server cluster, or to a Spark enabled Hadoop cluster running MATLAB Distributed Computing Server. For more information, see

- Big Data Workflow Using Tall Arrays and Datastores
- Use Tall Arrays on a Parallel Pool
- Use Tall Arrays on a Spark Enabled Hadoop Cluster

Parallel Menu Enhancement: Use the new menu items in the Parallel Menu to configure and manage cloud based resources

Open the Cloud Center web application and view MATLAB Distributed Computing Server license usage. For more information, see Use Parallel Menu and Cluster Profiles.

New Data Types in Distributed Arrays: Use enhanced functions for creating distributed arrays of: datetime; duration; calendarDuration; string; categorical; and table

Distributed calendarDuration Arrays:

calendarDuration calguarters cellstr time

caldays calweeks datevec

calmonths calyears iscalendarduration

Distributed categorical Arrays:

categorical iscategorical isundefined setcats

addcats iscategory removecats categories isordinal renamecats countcats isprotected reordercats

Distributed datetime Arrays:

exceltime datetime isweekend string between juliandate timeofday hms cellstr hour minute tzoffset datenum isbetween week month dateshift isdatetime posixtime year datevec isdst quarter ymd

day isnat second yyyymmdd

Distributed duration Arrays:

durationdatevechoursminutescellstrdaysisdurationsecondsdatenumhmsmillisecondsyears

Distributed string Arrays:

insertBefore replaceBetween string erase cellstr eraseBetween ismissing reverse extractAfter startsWith compose isstring contains extractBefore lower strip count extractBetween strlength pad endsWith insertAfter replace upper

Distributed tables:

table istable table2cell

head standardizeMissing tail

ismissing table2array

For more information, see Using MATLAB Functions on Distributed Arrays.

Loading Distributed Arrays: Load distributed arrays in parallel using datastore

Create distributed arrays more easily using datastore, and eliminate the need to create distributed arrays with codistributed. For more information, see Load Distributed Arrays in Parallel Using datastore.

datetime Support for Timestamps: Use built-in datetime objects in MATLAB to access timestamp information for jobs and tasks

You can now use the following properties for MATLAB jobs:

CreateDateTime
SubmitDateTime
StartDateTime

FinishDateTime

You can use the following properties for MATLAB tasks:

CreateDateTime
StartDateTime
FinishDateTime

For more information, see parallel. Job and parallel. Task

Data Transfer Measurement: Use ticBytes and tocBytes to measure the data transfer between MATLAB workers in a parallel pool

You can now measure how much data needs to be passed around to carry out parfor, spmd or parfeval. Use ticBytes and tocBytes to optimize your code and pass around less data.

Multithreaded Workers: Use multiple computational threads on your MATLAB workers

MATLAB workers used to run in single-threaded mode. Now you can control the number of computational threads so that workers can run in multithreaded mode and use all the cores on your cluster. This enables you to increase the number of computational threads, NumThreads, on each worker, without increasing the number of workers, NumWorkers. If you have more cores available, increase NumThreads to take full advantage of the built-in parallelism provided by the multithreaded nature of many of the underlying MATLAB libraries. For more information, see Create and Modify Cluster Profiles.

MathWorks Hosted License Manager: Use new option in MATLAB Distributed Computing Server script to ensure workers use MathWorks Hosted License Manager

You can now use a new -usemhlm option in your mdce script to ensure workers use MathWorks Hosted License Manager. For more information, see mdce.

JSON support for nodestatus: View the output of the nodestatus script in JavaScript Object Notation (JSON) format

You can now use a new -json option in your nodestatus script to view your output in JavaScript Object Notation (JSON) format. For more information, see nodestatus.

Upgrade Parallel Computing Products Together

This version of MATLAB Distributed Computing Server software is accompanied by a corresponding new version of Parallel Computing Toolbox software.

Compatibility Considerations

If you are running MATLAB Job Scheduler on your cluster, then you can upgrade MATLAB® Distributed Computing Server without upgrading Parallel Computing Toolbox. However, you cannot upgrade Parallel Computing Toolbox without upgrading MATLAB Distributed Computing Server.

If you are not running MATLAB Job Scheduler, then you must upgrade Parallel Computing Toolbox and MATLAB Distributed Computing Server together. Jobs created in one version of Parallel Computing Toolbox software will not run in a different version of MATLAB Distributed Computing Server software, and might not be readable in different versions of the toolbox software. The job data stored in the folder identified by JobStorageLocation (formerly DataLocation) might not be compatible between different versions of MATLAB Distributed Computing Server. Therefore, JobStorageLocation should not be shared by parallel computing products running different versions, and each version on your cluster should have its own JobStorageLocation.

Functionality Being Removed or Changed

Functionality	Result		Compatibility Considerations
Support for running MATLAB MapReduce on Hadoop 1.x clusters will be removed in a future release.		Hadoop 2.x or higher	Migrate MATLAB MapReduce code that runs on Hadoop 1.x to Hadoop 2.x

For more information, see Configure a Hadoop Cluster.

R2016a

Version: 6.8

New Features

Bug Fixes

Support for Distributed Arrays: Use enhanced distributed array functions including sparse input to direct (mldivide) and iterative solvers (cgs and pcg)

The following functions are new in supporting distributed arrays:

cgs conv conv2 expint ischar pcg superiorfloat

For more details, see MATLAB Functions on Distributed and Codistributed Arrays (Parallel Computing Toolbox).

In addition, the following features are new in providing enhanced functionality for distributed arrays:

- Sparse input to mldivide (direct system solve)
- Sparse input to cgs and pcg (iterative system solve)
- Single argument syntax for sprand and sprandn

Hadoop Kerberos Support: Improved support for Hadoop in a Kerberos authenticated environment

In R2016a, enhanced support for the recommended setup for the Cloudera distribution of Hadoop has been provided.

Transfer unlimited data between client and workers, and attached files up to 4GB in total, in any job using a MATLAB Job Scheduler cluster

In previous releases, the data transfer limit for a MATLAB Job Scheduler cluster was 2GB. In R2016a, this limit has been removed.

Third Party Scheduler Integration: Obtain integration scripts for Third Party Schedulers (IBM Platform LSF, Grid Engine, PBS and SLURM) from MATLAB Central File Exchange instead of Parallel Computing Toolbox

Obtain integration scripts for Third Party Schedulers (IBM® Platform LSF, Grid Engine family, PBS family and SLURM) from MATLAB Central File Exchange instead of Parallel Computing Toolbox.

- IBM Platform LSF
- Grid Engine family
- PBS family
- SLURM

Compatibility Considerations

In previous releases, integration scripts for Third Party Schedulers were available from Parallel Computing Toolbox. In R2016a, these integration scripts are obtained from MATLAB Central File Exchange and are no longer available from Parallel Computing Toolbox.

Upgrade parallel computing products together

This version of MATLAB Distributed Computing Server software is accompanied by a corresponding new version of Parallel Computing Toolbox software.

Compatibility Considerations

As with every new release, if you are using both parallel computing products, you must upgrade Parallel Computing Toolbox and MATLAB Distributed Computing Server together. These products must be the same version to interact properly with each other.

Jobs created in one version of Parallel Computing Toolbox software will not run in a different version of MATLAB Distributed Computing Server software, and might not be readable in different versions of the toolbox software. The job data stored in the folder identified by JobStorageLocation (formerly DataLocation) might not be compatible between different versions of MATLAB Distributed Computing Server. Therefore, JobStorageLocation should not be shared by parallel computing products running different versions, and each version on your cluster should have its own JobStorageLocation.

R2015b

Version: 6.7

New Features

Bug Fixes

Discontinued support for parallel computing products on 32-bit Windows operating systems

This release of MATLAB products no longer supports 32-bit Parallel Computing Toolbox and MATLAB Distributed Computing Server on Windows operating systems.

Compatibility Considerations

You can no longer install the parallel computing products on 32-bit Windows operating systems. If you must use Windows operating systems for the parallel computing products, upgrade to 64-bit MATLAB products on a 64-bit operating system.

Scheduler integration scripts for SLURM

This release offers a new set of scripts containing submit and decode functions to support Simple Linux Utility for Resource Management (SLURM), using the generic scheduler interface. The pertinent code files are in the folder:

matlabroot/toolbox/distcomp/examples/integration/slurm

where matlabroot is your installation location.

The slurm folder contains a README file of instructions, and folders for shared, nonshared, and remoteSubmission network configurations.

For more information, view the files in the appropriate folders. See also Program Independent Jobs for a Generic Scheduler and Program Communicating Jobs for a Generic Scheduler.

Improved performance of mapreduce on Hadoop 2 clusters

The performance of mapreduce running on a Hadoop 2.x cluster with MATLAB Distributed Computing Server is improved in this release for large input data.

parallel.pool.Constant function to create constant data on parallel pool workers, accessible within parallel language constructs such as parfor and parfeval

A new parallel.pool.Constant function allows you to define a constant whose value can be accessed by multiple parfor-loops or other parallel language constructs (e.g., spmd or parfeval) without the need to transfer the data multiple times.

For more information and examples, see parallel.pool.Constant.

Upgrade parallel computing products together

This version of MATLAB Distributed Computing Server software is accompanied by a corresponding new version of Parallel Computing Toolbox software.

Compatibility Considerations

As with every new release, if you are using both parallel computing products, you must upgrade Parallel Computing Toolbox and MATLAB Distributed Computing Server together. These products must be the same version to interact properly with each other.

Jobs created in one version of Parallel Computing Toolbox software will not run in a different version of MATLAB Distributed Computing Server software, and might not be readable in different versions of the toolbox software. The job data stored in the folder identified by JobStorageLocation (formerly DataLocation) might not be compatible between different versions of MATLAB Distributed Computing Server. Therefore, JobStorageLocation should not be shared by parallel computing products running different versions, and each version on your cluster should have its own JobStorageLocation.

R2015a

Version: 6.6

New Features

Bug Fixes

Support for mapreduce function on any cluster that supports parallel pools

You can now run parallel mapreduce on any cluster that supports a parallel pool.

Using DNS for cluster discovery

In addition to multicast, the discover cluster functionality of Parallel Computing Toolbox can now use DNS to locate MATLAB Job Scheduler clusters. For information about cluster discovery, see Discover Clusters. For information about configuring and verifying the required DNS SRV record on your network, see DNS SRV Record.

MS-MPI support for MATLAB Job Scheduler clusters

On 64-bit Windows platforms, Microsoft MPI (MS-MPI) is now the default MPI implementation for local clusters on the client machine.

For MATLAB Job Scheduler clusters on Windows platforms, you can use MS-MPI by specifying the -useMSMPI flag with the startjobmanager command.

Ports and sockets in mdce def file

The following parameters are new to the mdce_def file for controlling the behavior of MATLAB Job Scheduler clusters.

- ALL_SERVER_SOCKETS_IN_CLUSTER This parameter controls whether all client connections are outbound, or if inbound connections are also allowed.
- JOBMANAGER_PEERSESSION_MIN_PORT, JOBMANAGER_PEERSESSION_MAX_PORT These parameters set the range of ports to use when ALL_SERVER_SOCKETS_IN_CLUSTER = true.
- WORKER_PARALLELPOOL_MIN_PORT, WORKER_PARALLELPOOL_MAX_PORT These parameters set the range of ports to use on worker machines for parallel pools.

For more information and default settings for these parameters, see the appropriate mdce_def file for your platform:

- matlabroot\toolbox\distcomp\bin\mdce def.bat (Windows)
- matlabroot/toolbox/distcomp/bin/mdce def.sh (UNIX)

Compatibility Considerations

By default in this release, ALL_SERVER_SOCKETS_IN_CLUSTER is true, which makes all connections outbound from the client. For pre-R2015a behavior, set its value to false, which also initiates a set of inbound connections to the client from the MATLAB Job Scheduler and workers.

Discontinued support for GPU devices on 32-bit Windows computers

This release no longer supports GPU devices on 32-bit Windows machines.

Compatibility Considerations

GPU devices on 32-bit Windows machines are not supported in this release. Instead, use GPU devices on 64-bit machines.

Discontinued support for parallel computing products on 32-bit Windows computers

In a future release, support will be removed for Parallel Computing Toolbox and MATLAB Distributed Computing Server on 32-bit Windows machines.

Compatibility Considerations

Parallel Computing Toolbox and MATLAB Distributed Computing Server are still supported on 32-bit Windows machines in this release, but parallel language commands can generate a warning. In a future release, support will be completely removed for these computers, at which time it will not be possible to install the parallel computing products on them.

R2014b

Version: 6.5

New Features

Bug Fixes

Data Analysis on Hadoop clusters using mapreduce

MATLAB Distributed Computing Server supports the use of Hadoop clusters for the execution environment of mapreduce applications. For more information, see:

- · Configure a Hadoop Cluster
- · Run mapreduce on a Hadoop Cluster

Additional MATLAB functions for distributed arrays, including fft2, fftn, ifft2, ifftn, cummax, cummin, and diff

The following functions now support distributed arrays with all forms of codistributor (1-D and 2DBC), or are enhanced in their support for this release:

besseli besselj besselk bessely	erf erfc erfcinv erfcx erfinv	isinteger islogical isnumeric median mode
beta	fft2	pol2cart
betainc	fftn	psi
betaincinv	gamma	rgb2hsv
betaln	gammainc	sph2cart
cart2pol	gammaincinv	std
cart2sph	gammaln	toeplitz
compan	hankel	trapz
corrcoef	hsv2rgb	unwrap
COV	ifft	vander
cummax	ifft2	var
cummin	ifftn	
diff	isfloat	

For a list of MATLAB functions that support distributed arrays, see MATLAB Functions on Distributed and Codistributed Arrays.

R2014a

Version: 6.4

New Features

Bug Fixes

Duplication of an existing job, containing some or all of its tasks

You can now duplicate job objects, allowing you to resubmit jobs that had finished or failed.

The syntax to duplicate a job is

```
newjob = recreate(oldjob)
```

where oldjob is an existing job object. The newjob object has all the same tasks and settable properties as oldjob, but receives a new ID. The old job can be in any state; the new job state is pending.

You can also specify which tasks from an existing independent job to include in the new job, based on the task IDs. For example:

```
newjob = recreate(oldjob, 'TaskID', [33:48]);
```

For more information, see the recreate reference page.

More MATLAB functions enhanced for distributed arrays

The following functions now support distributed arrays with all forms of codistributor (1-D and 2DBC), or are enhanced in their support for this release:

eye ifft randi rand randn

Note the following enhancements for some of these functions:

- ifft and randi are new in support of distributed and codistributed arrays.
- rand(____, 'like',D) returns a distributed or codistributed array of random values of the same underlying class as the distributed or codistributed array D. This enhancement also applies to randi, randn, and eye.

For a list of MATLAB functions that support distributed arrays, see MATLAB Functions on Distributed and Codistributed Arrays.

Old Programming Interface Removed

The programming interface characterized by distributed jobs and parallel jobs has been removed. This old interface used functions such as findResource, createParallelJob, getAllOutputArguments, dfeval, etc.

Compatibility Considerations

The functions of the old programming interface now generate errors. You must migrate your code to the interface described in the R2012a Parallel Computing Toolbox release topic New Programming Interface.

matlabpool Function Being Removed

The matlabpool function is being removed.

Compatibility Considerations

Calling matlabpool continues to work in this release, but now generates a warning. You should instead use parpool to create a parallel pool.

R2013b

Version: 6.3

New Features

Bug Fixes

parpool: New command-line interface (replaces matlabpool), desktop indicator, and preferences for easier interaction with a parallel pool of MATLAB workers

This release introduces a number of enhancements for interacting with parallel pool resources. For more detailed descriptions of these enhancements, see parpool: New command-line interface (replaces matlabpool), desktop indicator, and preferences for easier interaction with a parallel pool of MATLAB workers in the Parallel Computing Toolbox release notes.

- Parallel pool syntax replaces MATLAB pool syntax for executing parallel language constructs such as parfor, spmd, Composite, and distributed. The pool is represented in MATLAB by a parallel.Pool object.
- A new icon at the lower-left corner of the desktop indicates the current pool status. Icon color and
 tool tips let you know if the pool is busy or ready, how large it is, and when it might time out. You
 can click the icon to start a pool, stop a pool, or access your parallel preferences.
- Your MATLAB preferences now include a group of settings for parallel preferences. These settings
 control general behavior of clusters and parallel pools for your MATLAB session. You can access
 your parallel preferences in the MATLAB toolstrip, from the parallel pool status icon, or by typing
 preferences at the command line.

For more information, see Parallel Preferences.

Compatibility Considerations

This release continues to support MATLAB pool language usage, but this support might discontinue in future releases. You should update your code as soon as possible to use parallel pool syntax instead.

Automatic start of a parallel pool when executing code that uses parfor or spmd

You can set your parallel preferences so that a parallel pool automatically starts whenever you execute a language construct that runs on a pool, such as parfor, spmd, Composite, distributed, parfeval, and parfevalOnAll.

Compatibility Considerations

The default preference setting is to automatically start a pool when a parallel language construct requires it. If you want to make sure a pool does not start automatically, you must change your parallel preference setting. You can also work around this by making sure to explicitly start a pool with parpool before encountering any code that needs a pool.

By default, a parallel pool will shut down *if idle for 30 minutes*. To prevent this, change the setting in your parallel preferences; or the pool indicator tool tip warns of an impending timeout and provides a link to extend it.

Option to start a parallel pool without using MPI

You now have the option to start a parallel pool on a local or MATLAB Job Scheduler cluster so that the pool does not support running SPMD constructs. This allows the parallel pool to keep running

even if one or more workers aborts during parfor execution. You explicitly disable SPMD support when starting the parallel pool by setting its 'SpmdEnabled' property false in the call to the parpool function. For example:

```
p = parpool('SpmdEnabled',false);
```

Compatibility Considerations

Running any code (including MathWorks toolbox code) that uses SPMD constructs, on a parallel pool that was created without SPMD support, will generate errors.

More MATLAB functions enabled for distributed arrays: permute, ipermute, and sortrows

The following functions now support distributed arrays with all forms of codistributor (1-D and 2DBC), or are enhanced in their support for this release:

ipermute	zeros
permute	ones
sortrows	nan
	inf
cast	true
	false

For more information about these functions and distributed arrays, see More MATLAB functions enabled for distributed arrays: permute, ipermute, and sortrows in the Parallel Computing Toolbox release notes.

Upgraded MPICH2 Version

The parallel computing products are now shipping MPICH2 version 1.4.1p1 on all platforms.

Compatibility Considerations

If you use your own MPI builds, you might need to create new builds compatible with this latest version, as described in Use Different MPI Builds on UNIX Systems.

Discontinued Support for parallel.cluster.Mpiexec

Support for clusters of type parallel.cluster.Mpiexec is being discontinued.

Compatibility Considerations

In R2013b, any use of parallel.cluster.Mpiexec clusters generates a warning. In a future release, support might be completely removed.

R2013a

Version: 6.2

New Features

Bug Fixes

Automatic detection and transfer of files required for execution in both batch and interactive workflows

Parallel Computing Toolbox can now automatically attach files to a job so that workers have the necessary code files for evaluating tasks. When you set a job object's AutoAttachFiles to true, an analysis determines what files on the client machine are necessary for the evaluation of your job, and those files are automatically attached to the job and sent to the worker machines.

For more information, see Automatic detection and transfer of files required for execution in both batch and interactive workflows in the Parallel Computing Toolbox release notes.

R2012b

Version: 6.1

New Features

Bug Fixes

Automatic detection and selection of specific GPUs on a cluster node when multiple GPUs are available on the node

When multiple workers run on a single compute node with multiple GPU devices, the devices are automatically divided up among the workers. If there are more workers than GPU devices on the node, multiple workers share the same GPU device. If you put a GPU device in 'exclusive' mode, only one worker uses that device. As in previous releases, you can change the device used by any particular worker with the gpuDevice function.

Detection of MATLAB Distributed Computing Server clusters that are available for connection from user desktops through Profile Manager

You can let MATLAB discover clusters for you. Use either of the following techniques to discover those clusters which are available for you to use:

- On the **Home** tab in the **Environment** section, click **Parallel > Discover Clusters**.
- In the Cluster Profile Manager, click Discover Clusters.

For more information, see Discover Clusters.