



# MATLAB - Parallelization and Performance

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## A word about Zoom

- Please stay muted unless necessary
- Questions: unmute, say "Question", remute, I will call you
  - -I will probably miss "Raise Hand" button
- Make sure microphone and screen sharing works
  - -Might be blocked by OS the first time
- Breakout rooms: groups of 2-3, do exercises collaboratively
  - -One person should share screen, rotate through





## **Round of introductions**

- Who are you?
- What is your research about?
- What do you use MATLAB for?
- How much experience do you have with MATLAB?





## Outline

- Part 1: Introduction
  - -Background
  - -MATLAB at Uni Siegen
  - -MATLAB parallel features
- Part 2: MATLAB parallelization
  - -Using MATLAB on the cluster
    - Exercise 1
  - -Parallel pools and cluster profiles
    - Exercise 2
  - Parallel programming in MATLAB
    - Exercise 3

- Part 3: Performance
  - -Profiling basics
  - -Serial performance
  - -Parallel performance
    - Exercise 4
  - -Mex functions
- Summary
  - Day 1/2 cut (approx.)





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## Background

- Lots of MATLAB users at Uni Siegen

   Sometimes very elaborate setups
- MATLAB not really built for HPC

   Rule of thumb: 1-2 orders of magnitude slower than C, Fortran
- However many features to support cluster use
- HPC community does not care much about MATLAB
- $\rightarrow$  What advice to give users?





## Background: idea

- Idea: approach problem by "simulating" a regular user
- Create own MATLAB application
  - -More complex than code examples/exercises
  - -Simpler than real scientific codes
- Parallelize application

   Note pitfalls, tips & tricks
- Optimize application
  - -Check against common HPC wisdoms
  - -Get a feeling for MATLAB performance





### Test code, basics

Code exists in multiple languages

-So far Python, MATLAB, Fortran, partially C++

- 3D panel method (fluid dynamics)
  - -Based on APAME (open source), which is based on VSAERO (NASA)
  - -Simplified physics compared to modern CFD
- Mix of various parallel techniques
  - Understand techniques
  - -Example for teaching purposes





#### Test code, equation system







#### Test code, data flowchart



MATLAB - Parallelization and Performance





#### Test code, operations

- Good mix of operations:
  - -Built-in mldivide for solving equation system
  - -Several built-in vecnorm function calls
  - -Some for loops
- Two compute-intensive phases:
  - -Fill B and C matrices
  - -Solve equation system
- Complexity roughly O(n^2)





#### Test code, data requirements

- Data object structure
  - -Constructor (allocates arrays)
  - -calc() method (fills arrays with values)
  - -Other methods as needed (called by calc() method)
- Total memory needed: (2 \* N^2 + 100 \* N ) \* 8 Byte
  - -More than 50 panels: matrices larger than everything else combined
  - -1000 panels  $\rightarrow$  16 MB
  - -10000 panels  $\rightarrow 1.6$  GB
  - -100000 panels  $\rightarrow$  160 GB





## Investigation method

- Implement
- Investigate performance
  - Is scaling logical?
  - Do common tips and tricks apply?
  - -What else did I notice?
- Parallelize with different strategies
  - -What is easy/hard to understand/implement
- Investigate parallel performance





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## MATLAB licensing at Uni Siegen

MATLAB available to everyone
 But not free for everyone

- Three types of licenses
  - Employee license
  - -Student license
  - Parallel server license

• No restriction on version





## MATLAB licensing: employees

• Available to all employees, managed by Fak. 4

• Large collection of toolboxes (including Parallel Computing)

May be installed on university or own computer
 May not be used on cluster outside Uni Siegen

Costs money (per user)
 Own PC and cluster = pay twice





## MATLAB licensing: students

- Students may use MATLAB for <u>free</u>
  - -Lectures, seminars etc.
  - -BSc/MSc thesis
  - -Even on cluster
- Same toolboxes as employee license
- May only be installed on own computer
- Not free: use for SHK work





## MATLAB licensing: cluster

• MATLAB Parallel Server

-Formerly: MATLAB Distributed Computing Server

- New: unlimited worker count
- Enables multi-node computations

   Multithreading still possible
- Parallel pool in SLURM job
  - Pools in general are part of Parallel Computing Toolbox





## **Getting MATLAB**

- ZIMT website
- Forms to fill out
- Detailed descriptions of license terms
- List of available toolboxes

https://www.zimt.uni-siegen.de/	eratung_und_lehre/software/?lang=de		
	Studienanfänger     Studierende     Beschäftigte		
ZIMT			
Aktuelles	Software-Dienste für die Universität Siegen		
Benutzerservice	Bestellformular (nur intern)		
IT-ServiceTeam Medien	Für die Angehörigen der Siegener Hochschule stellt das Zentrum für Informations- und Med Anbieter. Die Angebote für Studierende sind ausgewiesen.		
Hardware	Lizenzen, die von anderen Einrichtungen der Universität Siegen betreut werden, sind aus C		
Software	Lizenzen über das ZIMT		
Schulung	> Adobe - Mietlizenzen		
Dienste			
Kontakt	AUTODESK Education Programm (fur Studierende und Mitarbeitende)		
	Sorel CCL Academic Programm		
	> Docoloc [Plagiatserkennung]		
	Endnote [Literaturverwaltung] f ür Forschung und Lehre		
	> IBM SPSS Modeler - Campus Edition		
	> IBM SPSS Statistics - NRW-Landeslizenz (für Studierende und Mitarbeitende)		
	LabVIEW Campuslizenz		
	» Mathematica		
	MathWorks TAH Campus		
	» MathWorks-Campuslizenz TAH Student Standard (für Studierende)		





## MATLAB at Uni Siegen: miscellaneous info

- MATLAB training courses
  - -Mech. Eng. lecture: "Introduction to Computer Science"
    - Contains MATLAB intro segment
  - -Many users at ENC
    - Other intro course?
- Getting help:
  - -MATLAB documentation free, extensive, also usage tips, tutorials
  - -Our cluster website explains integration, downloads





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## Using MATLAB on the cluster: parallelism

- MATLAB on a cluster only makes sense if one uses parallelism
  - -More small computations
  - -Larger computations
  - -Same computation in shorter time
- MATLAB parallel features:
  - -Built-in multithreading
  - Parallel Toolbox
  - -MATLAB Parallel Server





## MATLAB multithreading

- Built into MATLAB
   Since at least R2014
- Does not require user action

   However does allow some control
- You have used it without noticing
- Most built-in functions use it

   Example: mldivide()
- Allows use of a full node (or PC), but only one





## MATLAB Parallel Toolbox

- Available as optional first-party toolbox
   Included in all Uni Siegen licenses
- Introduces concept of a "<u>worker</u>" —MATLAB process running in background
- Concept of a "parallel pool"
   Group of worker processes
- Large arsenal of functions to distribute work over workers





## MATLAB Parallel Toolbox

- MATLAB <u>client</u>: main MATLAB instance — Interactive (or command line)
- MATLAB <u>workers</u>: separate processes

   Interaction only through MATLAB
   code
- Worker <u>pool</u>: all workers

   Only one pool active for each client







### **MATLAB** Parallel Server

- Server-(i.e. cluster-)side component
   User only needs Parallel Toolbox
- Allows starting a pool on multiple nodes

   Can talk to SLURM
- Separate license
- Introduces concept of "cluster profile"
  - "Local" profile identical to running workers on same PC
  - -ZIMT provides "OMNI" profile





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## Using MATLAB on the cluster



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Performance





#### Using MATLAB on the cluster







## Using MATLAB on the cluster

• Different ways of using MATLAB

• Each has strengths and weaknesses

• Each requires varying levels of setup

Five scenarios on following slides, with advantages and disadvantages





## Using MATLAB: Scenario 1a

• Your computation is one M-file

- If you need more compute power, switch to cluster
  - –Copy M-file to cluster manually
  - -Create job script yourself
  - -Queue job like any other job







## Simple basic job script

• Simple job script

• Calls single MATLAB M script

- Script may open parallel pool internally
  - -Needs different job settings

1	#!/bin/bash
2	#SBATCH -p short
3	#SBATCH -t 0:01:00
4	#SBATCH nodes 1
5	#SBATCHntasks-per-node 1
6	<b>#SBATCH</b> cpus-per-task 64
7	#SBATCHmem 240000
8	
9	module load matlab
10	
11	matlab –nodisplay –r "demo1"





## Scenario 1a: comparison

#### Advantages:

- Very simple job script
- Complete control over SLURM
- Easily allows other operations outside MATLAB
- Local parpool on node(s) still possible...
- Can use own PC for everything except extensive calculations

#### Disadvantages:

- No interactivity
- Manual file transfer necessary
- Manual job setup and queuing necessary
- But no multi-node jobs
- Heterogenous systems, toolboxes between cluster and PC





## Using MATLAB: Scenario 1b

Your office Cluster • Like scenario 1a, but partially Login node (x2) SLURM Compile automated User File handling Compute nodes schedules Pre-process Run jobs Jobs Submit jobs Monitor jobs operates Script developed by ZIMT (SSH) 55 <u>PC</u> Test Prepare Ethernet (regular) Automated steps: Compile Infiniband (fast) SCP -Copying files to cluster File system File system Regular (/home etc.) Fast (Workspaces) -Generating job script -Queuing job GU Retrieving files after completion omni.m Built-in documentation • (+Workers) Demo 1





## Scenario 1b: comparison

#### Advantages:

- Highly automated
- Easy to use script
- Handles generation of parpool inside job if necessary
- Handles file synchronization between PC and cluster

#### Disadvantages:

- Self-built solution (may contain bugs)
- Hard-coded job script, not flexible
- Still no interactivity inside job
- Still heterogeneity between PC and cluster
- Requires password-less login on Windows





## Scenario 1b: script download

#### Script omni.m:

- Download from <u>https://cluster.uni-</u> <u>siegen.de</u>
- Can be adapted to other Slurm clusters
- Simplifies work

			Alle Martin		
	Getting Started	Access	Queuing a Job		
General Information	Wiss, Rechneri - Downloads		German		
Home	Downloads				
News					
Getting Started	Title	Categories	Publishing Date		
FAQ					
Events	Slides Cluster Introduction (English) 1 @ 4 downloads	, Winter 21/22 Schulungs	Schulungs 28. October 2021		
Local Training Courses		entertagen			
HPC Calendar	Slides from workshop "OMNI - New	Features and Changes"	30. September 2021		
Contact	PDF 1 (1) 16 downloads				
Downloads	Slides Linux Introduction Summer 2	021	16 July 2021		
media	PDF 1 ③ 142 downloads		10. 500 2021		
	Folien MATLAB-Einführung SoSe 202	1	14 July 2021		
OMNI Information	PDF 1 (1) 40 downloads	15	14. July 2021		
System Information	amai m				
Usage	1      58 downloads	MATLAB	24. June 2021		
File Systems					




#### Using MATLAB: Scenario 2a

- Connect to cluster via regular SSH
- Use MATLAB GUI on login node

   Edit scripts, brief test runs
   DO NOT RUN ACTUAL
   COMPUTATION ON FRONT-END
- Start pool from within MATLAB
   With OMNI profile
- Do compute-intensive work with pool (using Parallel Toolbox commands)







#### Scenario 2a: comparison

#### Advantages:

- Needs no MATLAB on PC
- Homogeneous between front-end and back-end
- Full interactivity (once job is running, you can do anything)
- No job script necessary: starting pool will automatically start job

#### Disadvantages:

- Ugly GUI
- Have to wait for job to start -Cannot run scripts
- Sometimes unstable SSH connection
   GUI sometimes does not launch
- Occasional MPI problems in the past
- Control over SLURM (e.g. queue) limited/complicated





#### Using MATLAB: Scenario 2b

- Connect to cluster and open MATLAB GUI

   Like 2a
- Do not use MATLAB to launch jobs, only to edit scripts, prepare job
- Launch jobs separately via sbatch

-Like method 1a







## Simple basic job script

#### Job settings:

- One task (one MATLAB client process)
  - More tasks if workers started
- CPUs per task determines MATLAB threads
- MATLAB called with -nodisplay

1	#!/bin/bash
2	#SBATCH -p short
3	#SBATCH -t 0:01:00
4	#SBATCHnodes 1
5	#SBATCHntasks-per-node 1
6	#SBATCHcpus-per-task 64
7	#SBATCHmem 240000
8	
9	module load matlab
10	
11	matlab –nodisplay –r "demo1"





#### Scenario 2b: comparison

#### Advantages:

- Combines advantages of 1a and 2a
   Flexibility
  - -Homogeneity
  - -Pool optional
- Can quickly look at job output files
- May use local pool on nodes (no extra wait time)
- Multiple jobs at the same time

#### Disadvantages:

- Also combines disadvantages
   Ugly GUI
  - -Manual file transfer
  - -Manual job setup
  - Worker pool either only local or second job
- Little interactivity with job





#### Using MATLAB: Scenario 3







### Scenario 3: comparison

#### Advantages:

- Compute resources reserved for you
- May still start pool with local profile
- Edit and run job on exact same MATLAB
- CLI MATLAB starts quickly, job continues on MATLAB crash

#### Disadvantages:

- Only CLI MATLAB available
- Initial job wait time
- Job dies if SSH connection to cluster closes





#### Not possible

- Cannot start pool on cluster from your PC
- Too much setup
- Not really supported by MATLAB devs
- We want you to think about resource usage







## Using a pool inside a job

- A job running MATLAB is still only one process
   Up to 64 cores via threads
- $\rightarrow$  Start a pool from inside job?
- Option A: use local profile
  - Up to 64 workers
  - -MATLAB client also running: 63 workers?
- Option B: use 'omni' profile, accept that second job is started –Multiple nodes possible (only if really needed please)
   Check resources: how much does MATLAB client need?





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- Exercises are very open
   Lots of time planned
- If you get done early, get creative
- Solve cooperatively (groups of three)
  - -One person shares screen
  - -Different person in each exercise
- I will come through





- Objectives
  - You can start MATLAB on the cluster
  - You understand the differences between the different approaches
- Tasks
  - -Connect to cluster
  - -Start MATLAB in GUI mode and in CLI mode (remember module load matlab)
  - -Start an interactive job with MATLAB

Note the following slide (before you begin the exercise)





- Notes
  - You have been given SLURM and Linux cheat sheets
  - You are allowed and indeed encouraged to use Google and the MATLAB documentation
  - Interactive jobs explained on cluster website: <u>https://cluster.uni-siegen.de/omni/usage/queuing-a-job/?lang=en#interactive</u>
- If bored, get creative
  - -Play around with SLURM job settings
  - -Open another Linux console with top, then run compute-intensive commands in MATLAB and watch CPU usage





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## MATLAB Multithreading

A word on maxNumCompThreads:

- Command to control number of threads
  - -Get: nThr = maxNumCompThreads
  - -Set: maxNumCompThreads (4)
    - Will return previous setting

• MATLAB documentation: "may be removed in future" (so far not)





### Creating clusters and pools

• Controlling parallel pool is done from inside MATLAB

• Either in GUI or programmatically

- Main actions:
  - -Add and edit cluster profiles
  - -Start and stop parallel pool
  - Send tasks to workers  $\rightarrow$  next section





### Cluster profile manager







#### Cluster profile manager Import button

•	antestar Profile Manager
० 👍 🔩 🛓	Buplicate & Rename & & Alarts Alerts Cloud Connectivity Properties see next slides
Discover Add Cluster Create Cloud Import Clusters Profile ▼ Cluster CREATE	Edit Delete + Set as Default Export Validate Cloud Center Hele
Cluster Profile GenericProfile1	horus Type: Slurm ( <u>How to configure</u> )
horus	Properties Variation
HPCServerProfile1	Stage Status Description
local (default)	🗹 Cluster connection test (parcluster) 🛛 🙎 Failed Unable to proceed because 'sinfo' is not on your path. Most likely this is
SlurmProfile 1	Skipped Validation skipped due to previous failure.
	SPMD job test (createCommunicatingJob) Ø Skipped Validation skipped due to previous failure.
	Pool job test (createCommunicatingJob) Ø Skipped Validation skipped due to previous failure.
	Parallel pool test (parpool) Skipped Validation skipped due to previous failure.
	Number of workers to use: Use default 🗘
	STACE DETAILS
	Stage started at 16:42:24. Completed in 0 min 2 cos
-	Stage started at 10.45.54. Completed in 0 min 2 sec.
'Local' profile	<b>Error Report:</b> Unable to proceed because 'sinfo' is not on your path. Most likely this is because your computer is not set up to submit to a Slurm cluster or because the Slurm scripts are not on your path.
always there	
Constants	Validate to see if
default	profile works
	Validate Show Report





- We have prepared a cluster profile for OMNI
  - -Download from website
  - Import into MATLAB GUI or CLI on cluster
- You might have to adjust some options
  - Some options might need adjusting on a case-by-case basis (e.g. walltime)
- Can be controlled programmatically (parcluster object)





- Pre-set and mostly selfexplanatory
  - -Leave these alone
- Maximum workers in a pool
  - -Can be at most 16
  - May still start pool with fewer workers
  - Can generally be left alone

	Description of this cluster Description	horus
	Folder where job data is stored on the client JobStorageLocation	current working folder (default)
	Number of workers available to cluster NumWorkers	16
T	Number of computational threads to use on each worker NumThreads	1
	Root folder of MATLAB installation for workers ClusterMatlabRoot	/cm/shared/mapps/matlab/2019a/
	License number (Optional: Used only if this cluster uses online licensing) LicenseNumber	<none></none>
•	Cluster uses online licensing RequiresOnlineLicensing	<none></none>





- Maximum threads per worker
  - -Pre-set to 1
  - -Depends on your use case
  - Emphasis on built-in functions: <=64, but fewer workers</p>
  - Emphasis on pool: 1 thread + 16 workers
  - Full power needed: 16 workers
     x 64 threads
    - 16-node job → long wait time

Description of this cluster Description	horus
Folder where job data is stored on the client JobStorageLocation	current working folder (default)
Number of workers available to cluster NumWorkers	16
Number of computational threads to use on each worker NumThreads	1
Root folder of MATLAB installation for workers ClusterMatlabRoot	/cm/shared/mapps/matlab/2019a/
License number (Optional: Used only if this cluster uses online licensing) LicenseNumber	<none></none>
Cluster uses online licensing RequiresOnlineLicensing	<none></none>





- Basic SLURM options
  - Pre-set to automatically start with correct number of tasks/cores
  - -Can be left alone
- Custom SLURM options
  - -Whatever you need
  - -Especially queue and walltime
  - Unfortunately has to be changed in profile every time

ADDITIONAL SLURM PROPERTIES	
Resource arguments for job submission. Use the placeholders '^N^' for the number of workers and '^T^' for NumThreads. ResourceTemplate	ntasks=^N^cpus-per-task=^T^ (default)
Additional command line arguments for job submission SubmitArguments	<none></none>
Script that cluster runs to start workers CommunicatingJobWrapper	MathWorks provided script (default)
FILES AND FOLDERS	
Automatically send code files to cluster. Data files must be listed below. AutoAttachFiles	true (default)
Manually specify files and folders to copy from client to cluster nodes (One entry per line) AttachedFiles	<none></none>
Manually specify folders to add to the workers' search path (One entry per line) AdditionalPaths	<none></none>





	ADDITIONAL SLURM PROPERTIES	
	Resource arguments for job submission. Use the placeholders 'ANA' for the number of workers and 'ATA' for NumThreads. ResourceTemplate	ntasks=^N^cpus-per-task=^T^ (default)
<ul> <li>Leave this alone</li> </ul>	Additional command line arguments for job submission SubmitArguments	<none></none>
	Script that cluster runs to start workers CommunicatingJobWrapper	MathWorks provided script (default)
<b>X</b>	Automatically send code files to cluster. Data files must be listed below. AutoAttachFiles	true (default)
Not sure what these do     Can be left alone	Manually specify files and folders to copy from client to cluster nodes (One entry per line) AttachedFiles	<none></none>
	Manually specify folders to add to the workers' search path (One entry per line) AdditionalPaths	<none></none>





- Limit number of workers
  - -Normally not needed
- Not sure how this works exactly

Range of number of workers to run job NumWorkersRange	[1 inf] (default)
Return command window output CaptureDiary	false (default)
Manually specify environment variables to copy from client to workers (One entry per line) EnvironmentVariables	<none></none>

 Might become important when using functionality external to MATLAB (MEX etc.)

# -Otherwise leave alone





#### Starting a pool from GUI







### MATLAB commands for pool control

- Everything discussed so far can be set with MATLAB commands
- Object-oriented:
  - Parallel pool object
  - -Cluster object
- Important functions:
  - -p = parpool() starts a new pool
    - Optionally specify profile or workers or both:
      - p = parpool(`omni',4)
  - -p = gcp() → get current pool
  - -c = parcluster('profile') → get cluster object
    - Edit SLURM settings: c1.SubmitArguments = <string>

Demo 5





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- Objectives
  - You understand the parallel pool settings
  - -You have functioning local and OMNI cluster profiles
- Tasks
  - -Open the cluster profile manager and examine the profile settings
  - -Import the omni\_m2020\_unlimited cluster profile (available at: <u>https://cluster.uni-siegen.de/omni/application-</u> <u>software/matlab/?lang=en#parallel-computations</u>
  - -Launch a pool with the GUI button
  - -Launch pools with different settings programmatically

#### Note the following slide (before you begin the exercise)





• Notes

– You should launch jobs in the short queue and with short walltimes to await waiting times

- If bored, get creative
  - -Examine parcluster and parpool object member variables
  - -Open another Linux console with top, then run compute-intensive commands in MATLAB and watch CPU usage
  - Examine the MATLAB "Parallel Preferences"

-...





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### Distributing tasks to workers

- Parallel functions (and associated concepts) in MATLAB

   Essentially mirrors usual parallel programming approaches
- Send task manually (not covered): batch(), parfeval()
- Parallel for loop: parfor
- Parallel program: spmd
- Interactive console on multiple workers: pmode
- Many MANY more
  - -E.g. Simulink: parsim/batchsim





#### The batch command

- batch command sends one task to one worker
- "Non-blocking": you can keep working on MATLAB client console
- Important concept: "future"
  - -Worker completes work
  - -Stores results
  - -Future object is how to get results back
- Job "diary": log file etc.





## Parallel for loops

- Very straightforward idea:
  - -Take a regular for loop
  - -Distribute loop iterations among workers
  - -Multiple iterations get done at the same time
- Simple to apply to existing code in principle...
- ...however there are some complications

   Loop iterations have to be 100% independent, many implications
- MATLAB documentation very helpful here





### Parallel for loops

• Basic parfor example

- Will run for a few seconds
- Note tic and toc for time measurement

https://www.mathworks.com/help/parallel-computing/parfor.html





### Parallel for loops: complications

- Five different types of variables
- MATLAB has to tell which variable is which type
  - -Does it exist once or is it copied on each worker?
  - -Does it change during the loop?
  - -What belongs to which worker?
- Error message if MATLAB cannot tell type
- Overall rule of thumb: the simpler the loop, the fewer problems
- Also useful for explaining parallel concepts in general





### Parallel for loops: variable types

1. Loop variable



https://www.mathworks.com/help/parallel-computing/troubleshoot-variables-in-parfor-loops.html




# Parallel for loops: loop variables

### 1. Loop variable

- Loop index
- Must be consecutive increasing integer
- Cannot break out of loop early
  - Loop iterations executed in parallel
  - Order not clear
- Cannot be modified

parfor	i	=	0:0.2:1
parfor	j	=	1:2:11
parfor	k	=	12:-1:1

- % not integers
  % not consecutive
- % not increasing

https://www.mathworks.com/help/parallel-computing/loop-variable.html





# Parallel for loops: sliced variables

- 2. <u>Sliced variable</u>
- Exists only once
- Distributed over workers
  - Each worker gets different slice
- Value may be changed inside loop

A = rand(1,10);
parfor ii = 1:10
 B(ii) = A(ii);
end

 MATLAB has to know how to distribute

https://www.mathworks.com/help/parallel-computing/sliced-variable.html





### Parallel for loops: broadcast variables

#### 3. Broadcast variable

- Existing variable before parfor block
- Copied to each worker
- Cannot be modified inside parfor
- Caution with large variables



https://www.mathworks.com/help/parallel-computing/broadcast-variable.html





### Parallel for loops: reduction variables

### 4. <u>Reduction variable</u>

- "Reduction": accumulate a value from each parallel task
  - Example: sum up value over each loop iteration
- Must be specific operation
- Cannot be dependent on iteration order

- X = X + expr
- X = X expr
- $X = X \cdot * expr$
- X = X \* expr
- X = X & expr
- $X = X \mid expr$
- X = [X, expr]
- X = [X; expr]
- X = min(X, expr)
- X = max(X, expr)
- X = union(X, expr)
- X = intersect(X, expr)

https://www.mathworks.com/help/parallel-computing/reduction-variable.html





### Parallel for loops: temporary variables

#### 5. Temporary variable

- Exists in each loop iteration separately
- Deleted at end of iteration
- If it exists before parfor, it is overwritten

```
a = 0;
z = 0;
r = rand(1,10);
parfor i = 1:10
    a = i; % Variable a is temporary
    z = z + i;
    if i <= 5
        d = 2*a; % Variable d is temporary
    end
end
```

https://www.mathworks.com/help/parallel-computing/temporary-variable.html





### Test code, data flowchart



MATLAB - Parallelization and Performance





## Code example: parallelizing a for loop

LES matrix is filled row by row in loop

 Rows independent: good candidate for parallelization

Influence data calculation (all panels on one point)
 Self-contained operation

• Result is copied into B and C matrices





### Code example: parallelizing a for loop

```
% calcMatsFor()
□ for iTarget=1:nTarget
                                                          % calcMatsParfor()
                                                         parfor (iTarget=1:nTarget,nWorkers)
                                                               xTarget = infl.panels.centers(iTarget,:);
     xTarget = infl.panels.centers(iTarget,:);
                                                               infl.calc( xTarget );
      infl.calc( xTarget );
     cMat(:,iTarget) = infl.cCoeffs;
                                                               cMat(:,iTarget) = infl.cCoeffs;
     bMat(:,iTarget) = infl.bCoeffs;
                                                               bMat(:,iTarget) = infl.bCoeffs;
                                                         - end
 - end
                                                          cMat = cMat.';
 cMat = cMat.':
                                                          bMat = bMat.';
 bMat = bMat.';
```

- Working parfor loop really possible with no changes
- Note how matrix is filled by column, then transposed





## MATLAB pmode: introduction

- Parallel mode (pmode) is a parallel interactive console
  - -Start with pmode start
  - -Will open new window and start pool
- Similar to regular MATLAB console
   But sends command to each worker
- Also understands constructs like codistributed arrays
- Extremely practical for understanding parallel concepts

   Also for debugging

Demo 6





### MATLAB spmd construct: background

- SPMD: general parallel programming term

   <u>Single Program Multiple Data</u>
- Program is written once, run multiple times in parallel
- Trivial case: each instance does exactly the same thing

   More common: each does same task for different part of input data
   However anything is possible
- Often includes communication between instances





### MATLAB spmd construct: introduction

- MATLAB: spmd block
  - -Block contents run on workers
- More complicated, but also more flexible than parfor
- Variables either copied or distributed over workers -Similar to parfor, but different mechanisms
- Important tools: labindex, numlabs
  - -Give ID of worker and total number of workers respectively
  - -Important to determine which worker works on what





### MATLAB spmd: codistributed arrays

- Important concept: codistributed array
  - -Array (or n-D matrix), treated mostly like regular array by MATLAB
  - -But each part lives on different worker
  - -User can say which part lives where
  - -Parallel programming term: Partitioned Global Address Space (PGAS)
- Distinction: distributed vs. <u>co</u>distributed array
  - -Distributed arrays not covered
- Also common: composite objects
  - -Object distributed over workers
  - -Not covered here





## MATLAB codistributed arrays

- SPMD construct requires moving data to workers manually
- Codistributor object: contains information on how array is distributed
  - -Can be along any axis (codistributor1d)
  - -Can also be block-cyclic (codistributor2dbc)
- Is otherwise treated like normal array
  - -Makes things easier for user
  - -Sometimes tricky to keep track which array is codist or not
- SPMD performance: need to make local copy

Demo 7





### MATLAB SPMD code example







### Second code example: solving the LES

- In addition to filling the LES, solving has to also be parallelized % calcDoublets doublets(:) = gather( mldivide(cCoeffs, rhs) ); % calcDoubletsParallel Even serial version needs = spmd modification (input data could cCoeffsDist = codistributed(cCoeffs); rhsDist = codistributed(rhs); be codistributed) solDist = mldivide(cCoeffsDist,rhsDist); Parallel solver needs to be doublets(:) = gather( solDist(:) ); inside spmd block
- Data needs to be scattered, then gathered again





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- Part 3: Performance
  - -Profiling basics
  - -Serial performance
  - -Parallel performance
    - Exercise 4
  - -Mex functions
- Summary





## **Exercise 3**

- Objectives
  - -You understand the parfor construct
  - -You understand the spmd construct and codistributed arrays
- Tasks
  - -Open pmode and create some codistributed arrays
  - -Implement the parfor example (Appendix A) and run it
  - -Implement the spmd example (Appendix B) and run it

Note the following slide





## Exercise 3

• Notes:

- You can leave out the commented code in the examples for now

- If bored, get creative
  - -Add new variables into parfor
  - Add new non-distributed, distributed and codistributed arrays to both examples

-...





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# Serial Performance, profiling

- Never optimize without measurement
- Simple stopwatch: tic and toc
  - -tic starts timer, toc ends it
  - -Can have multiple timers
- MATLAB profiler:
  - -Simply use profile on and profile off
  - -How often was each function called, how much time was spent in it
  - -Save HTML files with profsave

Demo 8

# Z(MT



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# Serial Performance

- How do I make MATLAB faster?
  - Throw out MATLAB and use a real programming language!

• Remember: never optimize without measurement!

- Common performance wisdoms in other languages:
  - -Minimize reallocation (e.g. C)
  - -Avoid explicit loops (e.g. Python)
  - -Use library functions (BLAS, LAPACK)
  - -Be aware whether the language is row-major or column-major





- Test: how much is the difference
  - 1. Minimize allocations, reallocations, copies
  - 2. Avoid explicit loops
  - 3. Use built-in functions
- Profile built-in function vs. explicit loops
   With and without reallocation
- Result:
  - -Built-in functions two orders of magnitude faster
  - Reallocation vs. no reallocation makes little difference

Demo 9





- Minimize allocations, reallocations, copies
  - -Difference less than expected
  - -That is not a good thing
- Avoid explicit loops
   Oh boy, yes!
- Use built-in functions
  - -Yes (because it's not MATLAB)

Best performance when using obvious way (that is a good thing)











- Right way round:
  - -load line once
  - -Calculate N results
  - -Unload once
- Wrong way round:
  - -Load line
  - -Only calculate one result
  - Unload line
  - –Load next line
  - -...
  - -Load first line again



### Column-major order







- Be aware whether language is row-major or column-major
  - Official: MATLAB is columnmajor (like Fortran)
  - -Test: flipped all indices
  - -Reality: "right way" is slower
    - Sometimes same time
    - Only exception: vecnorm() faster

### Row-major order









### Serial Performance, net vs. gross memory

- What is memory use of a Matlab array?
  - -Number of entries times 8 byte (double precision float)
  - -Plus overhead
  - -So what is overhead?

- Gross memory usage: Matlab will not tell you, but OS will
  - -Mac OS and Linux: 1.5 2 GB when idle
  - -Rough approximation





## Serial Performance, other stuff

- n threads never means n times the performance
  - -Overhead
  - -General problem in HPC

- Best friend for debugging: maxNumCompThreads(n)
  - -Sets number of threads to use
  - -Default same as CPUs available





# Serial Performance, panel3d Code

- Run panel3d code with no parallelization
- Measured on HoRUS, 1 node
- Varied N
- Varied maxNumCompThreads
- Recorded compute time with tic and toc
- Read CPU and RAM use from  ${\tt top}$





## Serial Performance, panel3d Code

Speedup panel3d with MATLAB. N=10000



May 17-18, 2021

MATLAB - Parallelization and Performance





# Serial Performance, panel3d Code

- Speed:
  - -Building up the matrices: barely scales with threads
    - Best performance at 6 cores (number of cores on 1 socket)
  - -Solve LES: scales almost perfectly (ScaLAPACK shipped with MATLAB)
- Memory
  - -LES buildup: memory overhead present but reasonable
    - CLI MATLAB has about 650 MB base RAM usage
    - Total measured: 2.4 GB RAM (theoretical 1.6 GB)
  - -LES solve: 3.2 GB RAM
    - Algorithm?

# Z(MT



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### Parallel performance

- Divide work up to workers
  - -Building LES
  - -Solving LES
- Try different strategies
  - -parfor
  - -spmd
  - -Local vs. Horus custer profile
- N = 10000, Number of workers 1, 2, 4, 8, 12, 16
- Determine speedup





# Speedup of matrix building operations

- Speedup has plateau
   Local profile: much earlier
- Most likely some unparallelizable operations
  - -Does not happen with N = 20000
- spmd-based operation performs better
  - Unclear cause







# Speedup of matrix building operations

- Starts to plateau with local profile
   Unclear cause
  - -Communication overhead of alg.?
- Considerably lower performance with parfor and Horus profile

   Strange: this is always an spmd, not a parfor
- 6 5 4 Speedup ~ parfor, local spmd, local -- parfor, horus 2 -- spmd, horus 1 0 0 4 8 12 16

nWorkers

Speedup solve, N=10000

- Reason:
  - -Automatic array distribution in parfor
  - -Needs to be redistributed




### Parallel Performance

• Both parfor and spmd accelerate calculation

• All less than n times serial

• spmd slightly faster

• Larger matrices possible with spmd -Likely due to distributed matrices





# mpiprofile

• At this point, just a shoutout to mpiprofile

• I have not seen such a comfortable tool for parallel performance measurement in any language

Demo 10





### spmd Lessons Learned

- Always keep track of whether a variable is distributed
- Assigning will overwrite

   Assigning like a(:) = b
   will give error when b is distributed
- Old Matlab rule about matrix operations rather than loops still holds

   Setting and accessing entries: prohibitively expensive
  - -Better: copy to local variable
- But: basically every Matlab function works with distributed arrays (!)

   Unsure which matrix distribution is best

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# **Exercise 4**

- Objectives
  - You can use the different profiling methods
  - You can interpret serial and parallel profiler outputs
- Tasks
  - -Add profiling commands in the spmd example (Appendix B) one by one, run the code and examine the output
    - $\bullet\, \mbox{tic}$  and toc
    - profile
    - mpiprofile

### Note the following slide (before you begin the exercise)





# Exercise 4

• Notes

-You can use the return command inside a script to exit it early

- If bored, get creative
  - -Add different functions and operation to the statements
  - -Start profiling your own MATLAB code
  - Implement and play around with the built-in function example (Appendix C)

-...

# Z(MT



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### What to do about performance

 Matrix calculation was tough to optimize – Limits of MATLAB reached?

• Idea: delegate matrix calculation function to non-MATLAB code

• MATLAB has interface for this

• MATLAB extension function: Mex





# Mex interface

- MATLAB has loads of interfaces
  - -Primarily to C and C++
  - -Also Python
  - -Also Fortran
  - -Both ways (run MATLAB from another language)
- Slightly confusing naming principle
  - -Matrix vs mex interface?
  - -Call libraries vs. call executables





# Mex interface

- To investigate:
  - -Fortran or C (or C++) interface more comfortable?
  - -How much programming knowledge needed?
  - -How well integrated is mex?
  - -How much better is performance
- What is mex exactly?
  - -Interface to system C, C++ or Fortran compiler (Linux: gcc)
  - -Stand-alone executable (outside MATLAB)
  - -Mex interface: list of C, C++ and Fortran functions to put in own code





# Interfacing Fortran code with MATLAB

- Strategy:
  - -Matrix calculation slow and badly parallelized
  - -panel3d Fortran version at least twice as fast as fastest MATLAB version
  - $\rightarrow$  Move calcMats() function to Fortran
- What needs to be given to Fortran
  - -Input: InflData object
  - -Output: B and C matrices
- Idea: implement this interface in Fortran (code exists)
  - -Fortran because matrix syntax similar to MATLAB
  - -Future: Also C and/or C++





# Interfacing Fortran code: approach

- Mex principle:
  - -Single Fortran routine called mexFunction() acts as entry/exit point
    - Use as wrapper for existing own code
  - Decode arguments

- What needs to be given to Fortran
  - -Input: InflData object
  - -Output: B and C matrices





# Interfacing Fortran code: own code

- Fortran code very similar to MATLAB
- Did not take long to convert
- Parallelization in this case similar to parfor logic
- Row-major vs. columnmajor more important





# Hello World mexFunction breakdown

- Single include 🛶
- Pointer to left-hand (output) and right-hand-side (input) args
   Number of args
  - -Requires checks for everything
- Pointer data type 🗕
- mex-specific library function

	implicit none
g	<pre>mwPointer :: plhs(*), prhs(*) integer :: nlhs, nrhs</pre>
	<pre>call mexPrintf( "Hello MATLAB world, this is Fortran!\n" )</pre>
e	subroutine mexFunction





# Decoding and re-encoding arguments

- This function gets one array from args
- Requires 4 steps
- Requires copying array

   Have not found better way
- Objects: repeat this for every member variable

	<pre>mwPointer, intent(in) :: ptr real(8), dimension(:,:), allocatable, intent(inout) :: mat</pre>
	integer :: nrows, ncols mwPointer :: dble_ptr !
	nrows = mxGetM( ptr ) ncols = mxGetN( ptr )
	allocate( mat( nrows, ncols ) )
	dble_ptr = mxGetDoubles(ptr)
	call mxCopyPtrToReal8(dble_ptr, mat )
nd	subroutine getDbleMatFromMatlabPtr

But: works, not complex if you understood it once





# Simplifying mex interface

- MATLAB File Exchange: Fortran 95 interface "with extras"
  - -<u>https://www.mathworks.com/matlabcentral/fileexchange/25934-fortran-</u> 95-interface-to-matlab-api-with-extras
  - -Community effort, not officially supported
  - -Have not tried yet
- Also: pointer interface is C logic
  - -Clearly, C support for mex came first
  - -More comfortable?
    - Problem: matrix handling in C less comfortable





### **MATLAB C++ Interfaces**

- Multiple C++ APIs, confusing naming
  - -C++ <u>mex</u> API: define C++ functions, call them from MATLAB
  - -C++ Data API: use MATLAB data structures in C++ code (arrays)
  - -C++ Engine API: talk to MATLAB from C++ (call functions)
  - -clibgen package: define interface to existing C++ library
  - $-\underline{C}$  API: separate, not covered
- Performance tested partially, <u>how comfortable is it to use</u>?





## **MATLAB C++ Interfaces**

- Advantages:
  - -C++ features very clean and well designed (understood it quickly)
  - -Much more comfortable for C++ programmers
  - -Object-oriented approach, including exceptions
- Disadvantages:
  - -C++ more strict about datatypes (Fortran too by the way)
  - -No functions like *sin*, *cos*, *sqrt* etc.
  - -You need to loop

#### Also, we still haven't looked at performance yet





### **MATLAB C++ Interfaces**

# Practical

MATLAB

Headers come with

namespaces

<pre>#include "mex.hpp"</pre>
<pre>#include "mexAdapter.hpp"</pre>
<pre>using matlab::engine::MATLABEngine;</pre>
using matlab::data::Array;
<pre>typedef matlab::data::TypedArray<double> DblArray;</double></pre>
<pre>typedef matlab::data::TypedArray<bool> BoolArray;</bool></pre>
<pre>typedef matlab::data::TypedArray<int64_t> Int32Array;</int64_t></pre>
<pre>class MexFunction : public matlab::mex::Function {</pre>
public:
<pre>void operator()(matlab::mex::ArgumentList outputs, matlab::mex::ArgumentList inputs)</pre>
{
<pre>std::cout &lt;&lt; "Hello MATLAB world, this is C++!" &lt;&lt; std::endl;</pre>
<pre>const Array&amp; inflIn = inputs[0];</pre>
<pre>const DblArray&amp; xTarget = inputs[1];</pre>
<pre>const Engine engine = getEngine();</pre>
InflData infl( engine, inflIn );
<pre>infl.calc(xTarget);</pre>
}
};





# Code comparison

MATLAB	<pre>function pVec = calcPVectors(centers,xTarget,pVec)     pVec(:,:) = xTarget - centers; end</pre>
Fortran	<pre>subroutine calcPVectors(centers,xTarget,p)     !</pre>
	<pre>real(8), dimension(:,:), intent(in) :: centers real(8), dimension(NDIMS), intent(in) :: xTarget real(8), dimension(size(centers,1),NDIMS), intent(inout) :: p </pre>
	integer :: iPanel !
	<pre>do iPanel = 1,size(centers,1)     p(iPanel,:) = xTarget(:) - centers(iPanel,:) end do</pre>
	end subroutine calcPVectors





# Code comparison



C++

static void calcPVectors( const DblArray& centers, const DblArray& xTarget, DblArray& pVec)





### MATLAB C++ Performance

• Test: simply recreate algorithm with MATLAB-provided Matrix classes

• Result: extremely slow

• Unfortunately, this complicates things

• Use raw pointers again?





# Calling mex

- Simple command line tool
  - -Already available on cluster
  - -Even cmake knows MATLAB and mex (!)
- One call:

mex -I../Release\_horus ../src/mex/calcMatsMex.F90

- Include directory with own code
  - Needs to be compiled separately
- Easy setup compared to usual C or Fortran compilers -Complexity is only in the way interface works





# Summary of mex experiments

- MATLAB code can potentially benefit greatly from external code
  - -Several times faster
  - -Complete arsenal of optimizations (vector registers etc.)
  - -May parallelize with OpenMP and MPI
- Fortran interface is extremely awkward —Clearly a C relic
- Getting mex to run is however very simple
  - -Already installed
  - -Minimal includes/library links etc.





### **Recommendations for mex use**

- Minimize interface interaction
  - -Few input/output arguments
  - -Simple types (arrays)

- Choose interface wisely
  - -Which language do you know better?

- Good candidates for delegation to other language
   Eulertion that takes some arrays, has hig loop(s), outputs some
  - -Function that takes some arrays, has big loop(s), outputs some other arrays

# Z(MT



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# Summary

- Wide variety of parallel options
  - -Comfortable profiling and debugging (mostly)
- Built-in functions often already optimized
  - Two orders of magnitude when written in another language
- Built-in multithreading
  - -Helps a lot
  - "Free" (no additional coding)
- parfor
  - -Easy to implement
  - -Not transparent





## Summary

- spmd
  - -Lots and lots of possibilities
  - -Complex
  - -Lots of potential for errors
- Scaling good for built-in functions, bad for own functions
- mex
  - -Initially tricky
  - Potentially huge benefit
  - -Fortran: code very similar





### Recommendations

- Best approach for writing your code:
  - 1. Use MATLAB for code until it works (rapid prototyping)
    - Focus on mathematical operations
    - Comfortable debugging
    - Built-in visualization
  - 2. Once it works, analyze performance





### Recommendations

- 3. Identify functions where most time is spent
  - Simple in programming terms (likely complex in mathematical terms)
  - Big long loops inside

- 4. Delegate these to mex function
  - Feasible(!) after initial complexity
  - Fortran: similar to MATLAB syntax, C closer to interface logic





### Thank you for your attention

**Questions?** 

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# Feedback round

- What were your expectations, and where they fulfilled?
- What was your favorite part about the course?
- What did you dislike or what do you feel can be improved?
- How did you learn about this course?
- What other topics would you like to see in future ZIMT courses?





# **Appendices: code examples**

- Appendix A: parfor example
- Appendix B: SPMD test example
- Appendix C: built-in functions performance example





### Appendix A: parfor example







# Appendix B: spmd example

```
n = 10000;
 maxNumCompThreads
 As = rand(n);
 rs = rand(n,1);
 tic
 xs = As \setminus rs;
 toc
₽ spmd
      codist1 = codistributor1d(1);
      Ap = rand(n,codist1);
      rp = rand(n,1,codist1);
      tic
      xp = Ap \setminus rp;
      toc
  end
```





### Appendix C: built-in functions performance example

clear;
n = 20000;
A = rand(n);
b = rand(n,1);
c = zeros(1,n);
% Without reallocating c
tic
c(:) = A * b;
toc
%% With possible reallocation of c
tic
c = A * b;
toc
%% With newly allocated result
tic
c2 = A * b;
toc
%% Manual matrix multiplication
tic
for i = 1:n
for j = 1:n
C(1) = C(1) + A(1, j) * D(j);
- end
toc
%% Manual matrix multiplication with reallocation in each step
tic
$c_{3(i)} = 0.0$
for $i = 1:n$
$c_3(i) = c_3(i) + A(i, j) * b(j);$
- end
- end
toc