INTENSIVE COMPUTATION

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Lecture 3

2018-2019

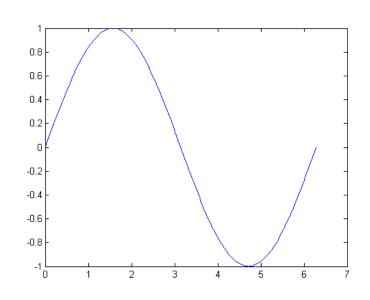
INTRODUCTION TO MATLAB

Part 2

The function plot creates a 2D line plot - it can be used in different ways

Example

```
» n = 31
» x = linspace(0,2*pi,n)
» y = sin(x)
» plot(x,y)
```



x is a vector of linearly spaced values between 0 and 2π

y is the vector of values of sine function evaluated at the values in x

• Command plot is:

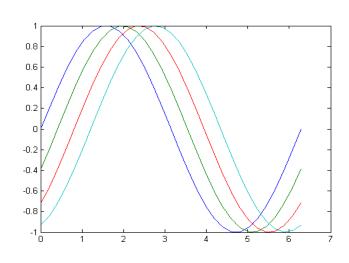
```
• plot(X,Y,options)
```

Where **x** is for abscissas and **y** is for ordinates options sets the *line style*, *marker symbol*, and *color*

To plot multiple lines in the same windows, we can use two ways:

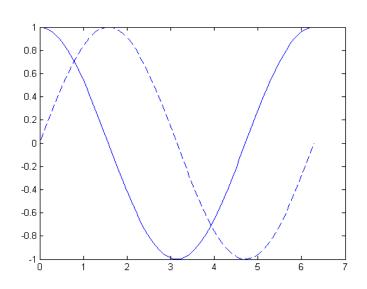
```
y2 = sin(x - .4);
y3 = sin(x - .8);
y4 = sin(x - 1.2);

• plot(x,y,x,y2,x,y3,x,y4)
• plot(x,[y;y2;y3;y4])
```



 Another way to plot multiple line in the same window is by using commands hold on and hold off:

```
» x = linspace(0,2*pi)
» y1 = cos(x)
» y2 = sin(x)
» plot(x,y1,'-')
» hold on
» plot(x,y2,'--')
» hold off
```



You can add a title and axis labels to the graph

```
» title('title of the graph')
» xlabel('x axis')
» ylabel('y axis')
```

- axis axis scaling and appearance
- legend graph legend
- text create text object in current axes
 » text(x(70)+0.5,r(70),'r = -2x')
- grid on add grid lines for 2D and 3D plots

Other functions for graphs are:

• loglog Log-log scale plot

• semilogarithmic plot (x logarithmic, y linear)

• **semilogy** Semilogarithmic plot (x linear, y logarithmic)

errorbar
 Plot error bars along curve

• bar Bar graph

• stairs Stairstep graph

• scatter Scatter plot

subplot divides the current figure into grid, it numbers the cells by rows

» subplot(m,n,p)

divides the current figure into an m-by-n grid and plots in the grid position specified by p

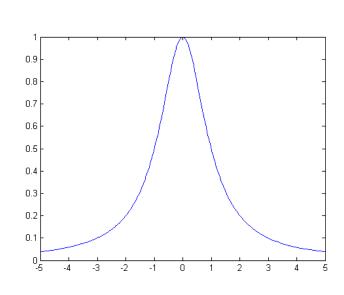
 1
 2

 3
 4

fplot(fun, lims) plots a function

- fun, that must be a string
- between the limits specified by lims, specifying the x-axis limits
 ([xmin xmax]), or the x- and y-axes limits, ([xmin xmax ymin ymax])

```
» fun='1/(1+x^2)';
» lims=[-5,5];
» fplot(fun,lims);
or the equivalent
» fplot('1/(1+x^2)', [-5,5]);
```



• fplot(fun,limits,LineSpec) plots fun using the line specification LineSpec

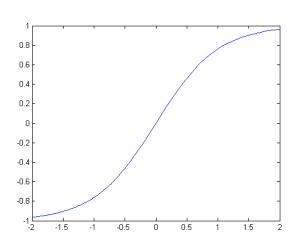
```
fplot(fun, lims, '- -')
fplot(fun, lims, 'r -')
```

• fplot can plot a vector of functions

```
fplot('[sin(t), sin(t-.25), sin(t-.5)]',[0,2*pi])
```

- **ezplot** plots the expression fun(x) over the default domain $-2\pi < x < 2\pi$, where fun(x) is an explicit function of only x
- ezplot(fun,[xmin,xmax]) plots fun(x) over the domain: xmin< x < xmax
- Both for fplot and ezplot fun can be a function handle

```
fh = @tanh;
fplot(fh,[-2,2])
```



3D plot with mesh and surf

- mesh and surf plot a surface
- mesh and surf create 3D surface plots of matrix data generated by the command meshgrid

```
» n=30; m=n;

» x=linspace(-2,2,n);

» y=linspace(-2,2,n);

» [X,Y]=meshgrid(x,y); % matrices X e Y for the grid

» Z=(1-Y).*cos(X.^2)+(X-1).*cos(Y.^2);
```

 \gg mesh(X,Y,Z);

You can load variables from file into workspace with load

For example if you want analyze data coming from a program, like the following, that are in the file data.dat

1	0.2000	-5
2	0.2500	-9
3	0.0740	-23
4	0.0310	-53
5	0.0160	-105
6	0.0090	-185
7	0.0050	-299
8	0.0030	-453
9	0.0020	-653
10	0.0020	-905

If you load these data with the function **load**, a matrix is created of size 10x3

```
>> load data.dat
>> whos
Name Size Bytes Class
data 10x3 240 double array
Grand total is 30 elements using 240 bytes
load filename is the command form
load 'filename' is the function form
```

```
>> M = load('data.dat')
M =
1.0000
       2.0000 -5.0000
2.0000
       0.2500 - 9.0000
3.0000 \quad 0.0740 \quad -23.0000
4.0000 0.0310 -53.0000
5.0000 0.0160 -105.0000
6.0000
      0.0090 -185.0000
7.0000
       0.0050 - 299.0000
8.0000 \quad 0.0030 \quad -453.0000
9.0000 0.0020 -653.0000
10.0000 0.0020 -905.0000
```

save save workspace variables to file

- save (filename)
 saves all variables from the current workspace in a formatted binary file (MAT-file) called filename
 if filename is not specified the file Matlab.mat is created
- save (filename, variables)
 saves only the variables or fields of a structure array specified by variables
- save (filename, variables, fmt)
 saves in the file format specified by fmt variables is optional

Example

```
% mytable.m
n=input('Insert the number of values n:');
x=linspace(0,pi,n);
s=sin(x);
c=cos(x);
v=(1:n);
save mytable.dat v x s c -ascii
```

Example

To visualize the table saved in the previous example with save we can load the file and display the table

```
% viewtable.m
load mytable.dat
A=mytable;
disp('-----');
fprintf('k\t x(k)\t sin(x(k))\t cos(x(k))\n');
disp('-----');
fprintf('%d\t %3.2f\t %8.5f\t %8.5f\n',A);
```

```
dir List directory
  dir directory_name or dir('directory_name') lists the
  files in a directory -- Pathnames and wildcards may be used
  dir *.m lists all the M-files in the current directory
D = dir('directory_name') returns the results in an M-by-1
```

D = dir('directory_name') returns the results in an M-by-1
structure with the fields:

```
name -- filename

date -- modification date

bytes -- number of bytes allocated to the file

isdir -- 1 if name is a directory and 0 if not

datenum -- modification date as a MATLAB serial date number
```

Techniques for Improving Performance

Preallocating Arrays

- for and while loops that incrementally increase the size of a data structure each time through the loop can adversely affect performance and memory use
- resizing arrays often requires MATLAB to spend extra time looking for larger contiguous blocks of memory, and then moving the array into those blocks
- you can improve code execution time by preallocating the maximum amount of space required for the array

Techniques for Improving Performance

- Preallocating a Nondouble Matrix
 - Use the following command to create an array having int8 values

- This command allows to save time and memory
- Avoid using the following method, when you preallocate a block of memory to hold a matrix of some type other than double

```
A = int8(zeros(100)) ← INEFFICIENT
```

• This statement preallocates a 100-by-100 matrix of int8, first by creating a full matrix of double values, and then by converts each element to int8

Techniques for Improving Performance

Vectorization

- MATLAB is optimized for operations involving matrices and vectors
- The process of revising loop-based, scalar-oriented code to use MATLAB matrix and vector operations is called *vectorization*
- Vectorizing your code is worthwhile for several reasons:
 - *Appearance*: Vectorized mathematical code appears more like the mathematical expressions, making the code easier to understand
 - Less Error Prone: Without loops, vectorized code is often shorter, and fewer lines of code mean fewer programming errors
 - Performance: Vectorized code often runs much faster

Vectorizing Code for General Computing

This code computes the sine of 1,001 values ranging from 0 to 10:

```
i = 0;
for t = 0:.01:10
  i = i + 1;
  y(i) = sin(t);
end
```

This is a vectorized version of the same code:

```
t = 0:.01:10;
y = sin(t);
```

Vectorizing Code for Specific Tasks

This code computes the cumulative sum of a vector at every fifth element:

```
x = 1:10000;
ylength = (length(x) - mod(length(x),5))/5;
y(1:ylength) = 0;
for n= 5:5:length(x)
  y(n/5) = sum(x(1:n));
end
```

This code shows one way to accomplish the task:

```
x = 1:10000;
xsums = cumsum(x);
y = xsums(5:5:length(x));
```

Array Operations

- Array operators perform the same operation for all elements in the data set
- Example
 - collect the volume (V) of various cones by recording their diameter (D) and height (H)
 - The volume for that single cone: v = 1/12*pi*(D^2)*H
 - Consider 10,000 cones
 - The vectors D and H each contain 10,000 elements

```
for n = 1:10000

V(n) = 1/12*pi*(D(n)^2)*H(n));
end
```

Vectorized Calculation

```
V = 1/12*pi*(D.^2).*H;
```

More examples

Use built-in Matlab functions

- find is a very important function
 - Returns indices of nonzero values
 - Can simplify code and help avoid loops
- Basic syntax: index=find(cond)

```
» x=rand(1,100);

» inds = find(x>0.4 & x<0.6);</pre>
```

- Inds will contain the indices at which x has values between 0.4 and 0.6.
- This is what happens:
 - x>0.4 returns a vector with 1 where true and 0 where false
 - x<0.6 returns a similar vector
 - The & combines the two vectors using an and
 - The find returns the indices of the 1's

More examples

Given x= sin(linspace(0,10*pi,100)), how many of the entries are positive?

```
    Using a loop and if/else
```

```
count=0;
for n=1:length(x)
  if x(n)>0
     count=count+1;
  end
end
```

Being more clever

```
count=length(find(x>0));
```

Avoid loops! Built-in functions will make it faster to write and execute

PARALLEL TOOLBOX

Parallel Computing

- Parallel Computing: Using multiple computer processing units (CPUs) at the same time to solve a problem
- The compute resources might be:
 - computer with multiple processors or
 - networked computers
- The computational problem should be able to:
 - Be broken into discrete parts that can be solved simultaneously and independently
 - Be solved in less time with multiple compute resources than with a single compute resource.

Parallel Computing in Matlab

- Parallel Computing Toolbox (PCT)
 - shared memory, single node
 - parfor
- Matlab Distributed Computing Server (MDCS)
 - distributed computing across nodes
 - spmd or parfor
- Built-in multithreading
 - shared memory, single node

Parallel Computing in MATLAB

- MATLAB Parallel Computing Toolbox
 - Workers limited only by resources on the node, see parallel preferences for the default. Typically the entire node.
 - Built in functions for parallel computing
 - parfor loop (for running task-parallel algorithms on multiple processors)
 - spmd (handles large datasets and data-parallel algorithms)

Primary Parallel Commands

- parpool
 - mypool = parpool(4)
 - ... do work ...
 - delete(mypool)
- parfor (for loop)
- spmd (distributed computing for datasets)

parpool

- Use parpool to open a pool of workers to execute code on other compute cores
- In Matlab you can think of workers like threads or processes
- You can open these workers locally (on the same node) or remotely
- Local access is enabled by the Parallel Computing Toolkit, remote access is enabled via MDCS (Matlab Distributed Computing Server)

parpool

Starting a parallel pool

```
• mypool = parpool ('local',4);
• Mypool = parpool(4);
```

- Opens 4 workers locally on the same node
- Communication is fastest within a node
- Make sure you submitted your Matlab job with "-n X" where X matches the number of workers you open! Use –N 1 to ensure the slots are all on the same node (i.e. local)
- Use display (mypool) to show information about the pool

Closing a parallel pool

- delete (mypool) to end parallel session
- If you didn't save the name you can use

```
• delete (gcp('nocreate'));
```

Parallel for Loops (parfor)

- parfor loop executes a series of statements in the loop body in parallel
- A parfor-loop can provide significantly better performance than its analogous for-loop, because several MATLAB workers can compute simultaneously on the same loop
- Each execution of the body of a parfor-loop is an iteration
- The MATLAB client issues the parfor command and coordinates with MATLAB workers to execute the loop iterations in parallel on the workers in a parallel pool
- The client sends the necessary data on which **parfor** operates to workers, where most of the computation is executed
- The results are sent back to the client and assembled

Parallel for Loops (parfor)

- MATLAB workers evaluate iterations in no particular order and independently of each other
- Because each iteration is independent, there is no guarantee that the iterations are synchronized in any way, nor is there any need for this
- If the number of workers is equal to the number of loop iterations, each worker performs one iteration of the loop
- If there are more iterations than workers, some workers perform more than one loop iteration; in this case, a worker might receive multiple iterations at once to reduce communication time

Parallel for Loops (parfor)

A parfor-loop can be useful if you have a slow for-loop.

Consider parfor if you have:

- Some loop iterations that take a long time to execute.
 - In this case, the workers can execute the long iterations simultaneously.
 - Make sure that the number of iterations exceeds the number of workers. Otherwise, you will not use all workers available.
- Many loop iterations of a simple calculation (such as a Monte Carlo simulation or a parameter sweep)
 - parfor divides the loop iterations into groups so that each worker executes some portion of the total number of iterations.

Parallel for Loops (parfor)

A parfor-loop might not be useful if you have:

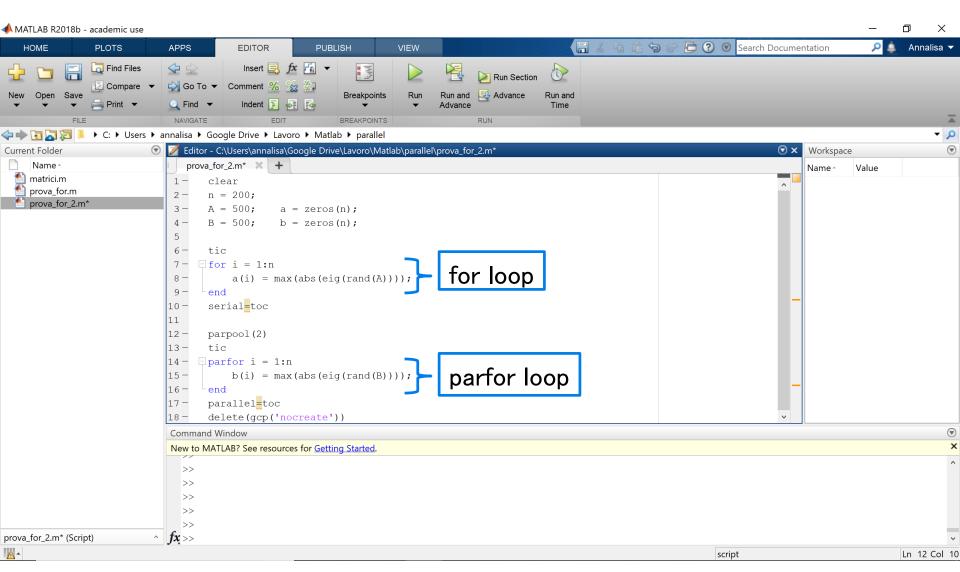
- Code that has vectorized out the for-loops
 - If you want to make code run faster, first try to vectorize it
 - Vectorizing code allows you to benefit from the built-in parallelism provided by the multithreaded nature of many of the underlying MATLAB libraries
 - However, if you have vectorized code and you have access only to local workers, then
 parfor-loops may run slower than for-loops
 - Do not devectorize code to allow for parfor; in general, this solution does not work well
- Loop iterations that take a short time to execute
 - In this case, parallel overhead dominates your calculation

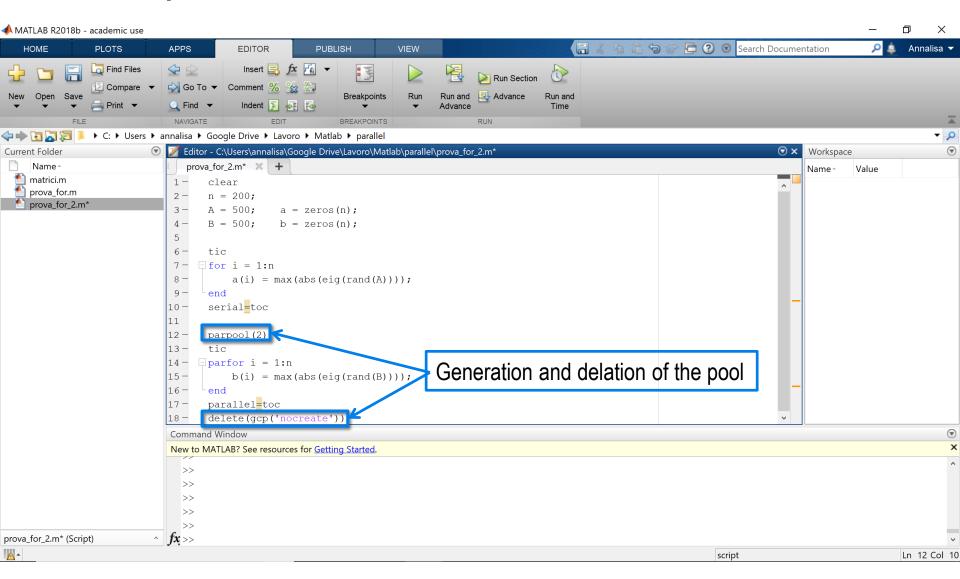
Parfor example

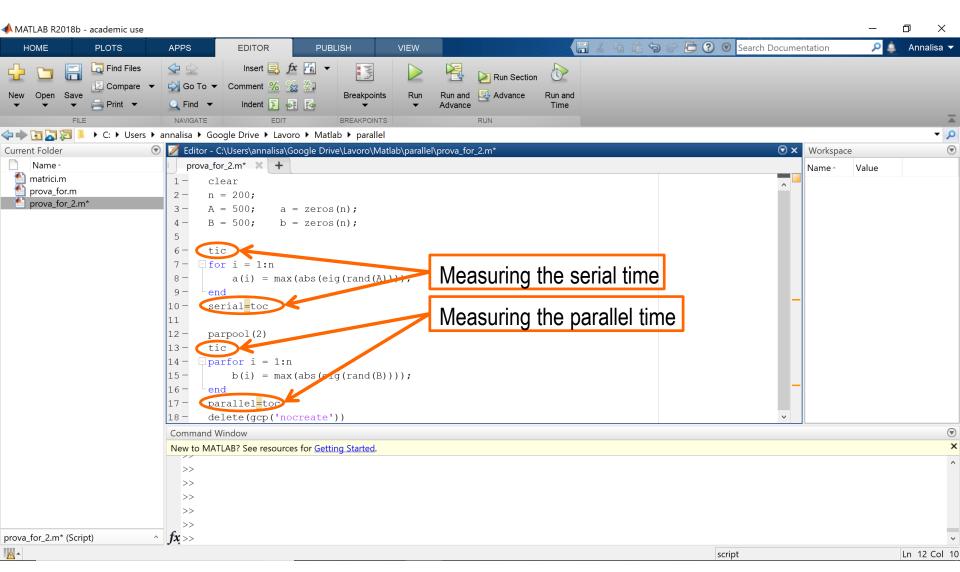
Will work in parallel, loop increments are not dependent on each other

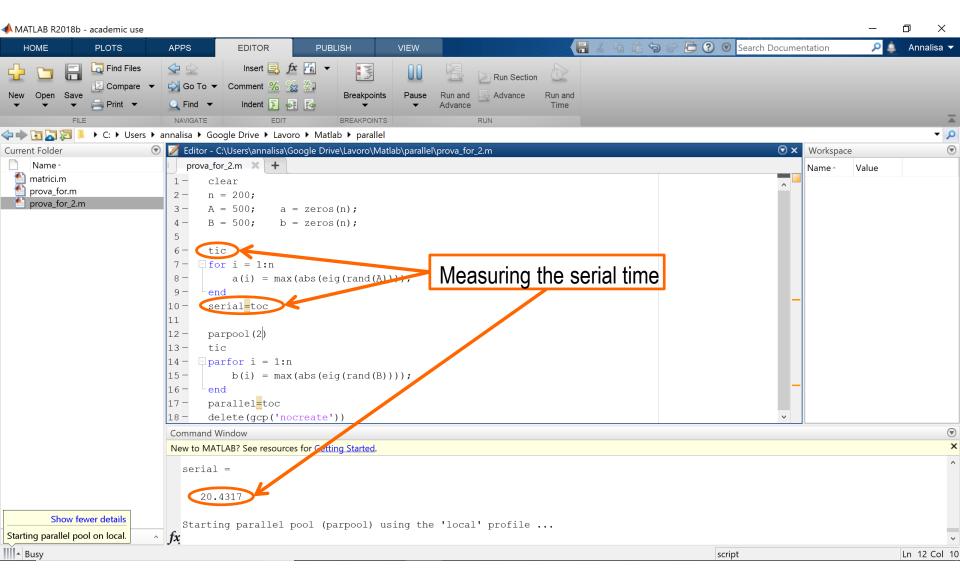
Serial Loop example

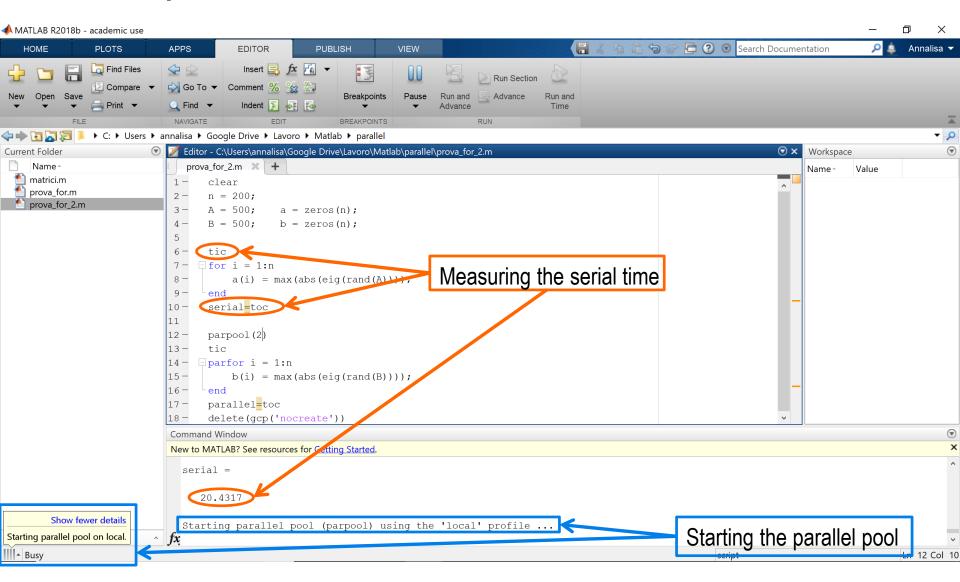
DOES NOT work in parallel - it's serial:

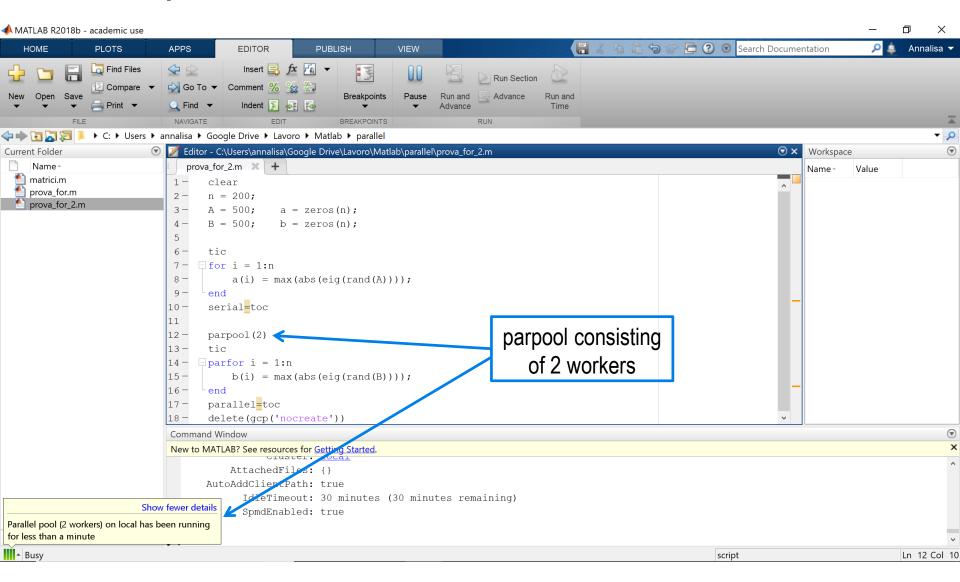


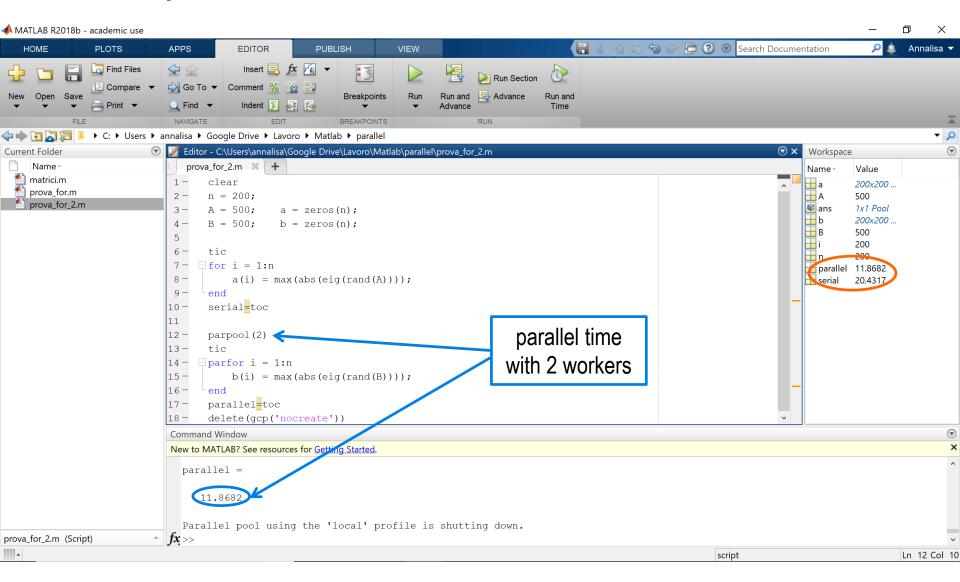


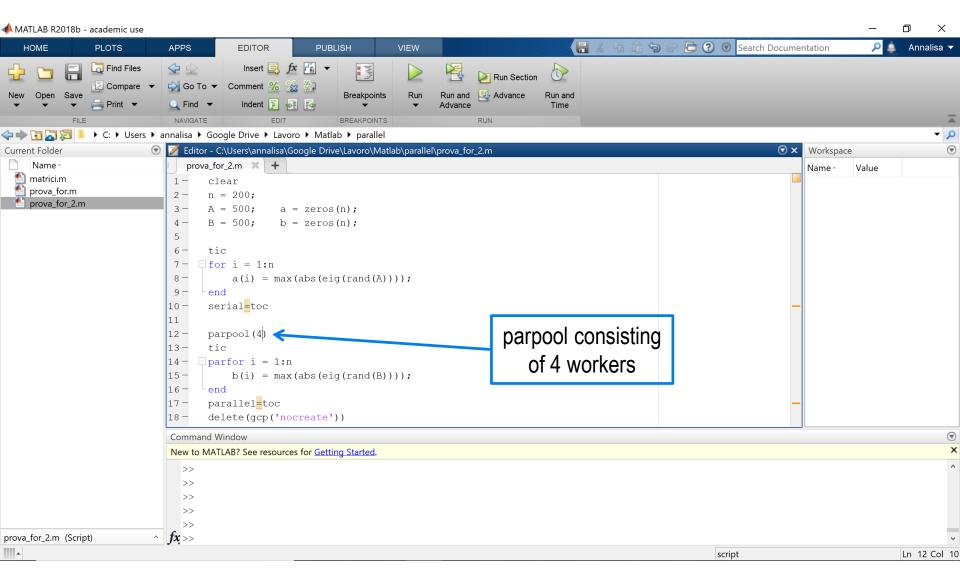


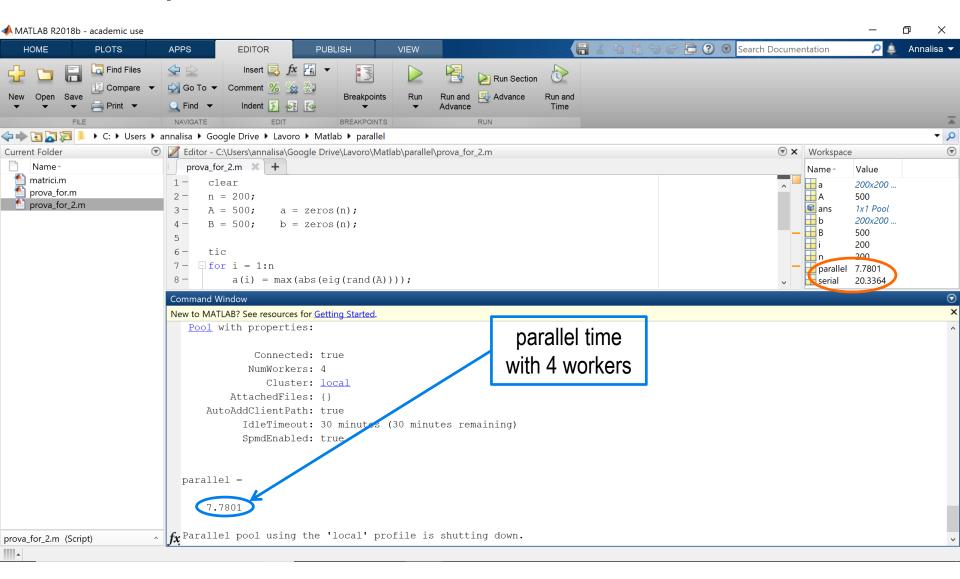


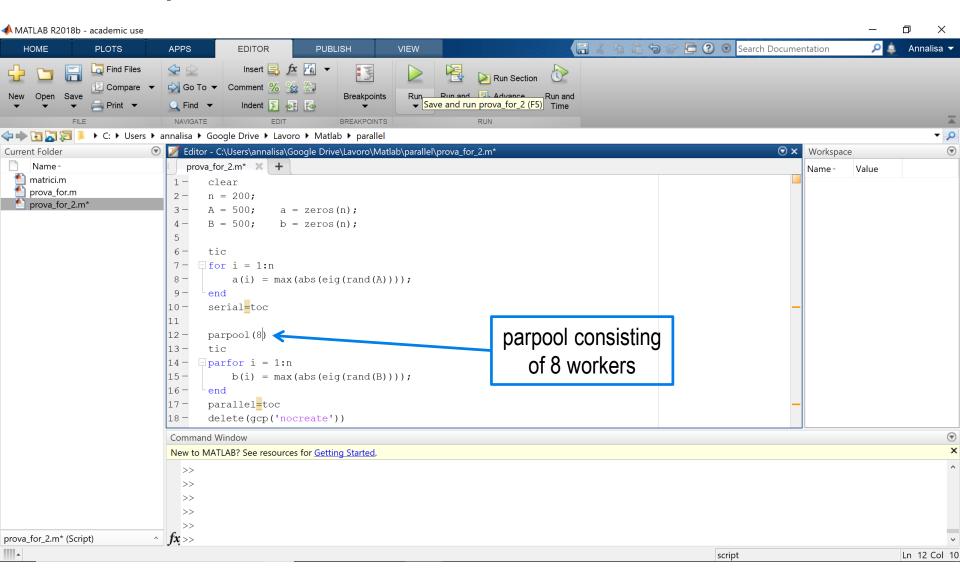


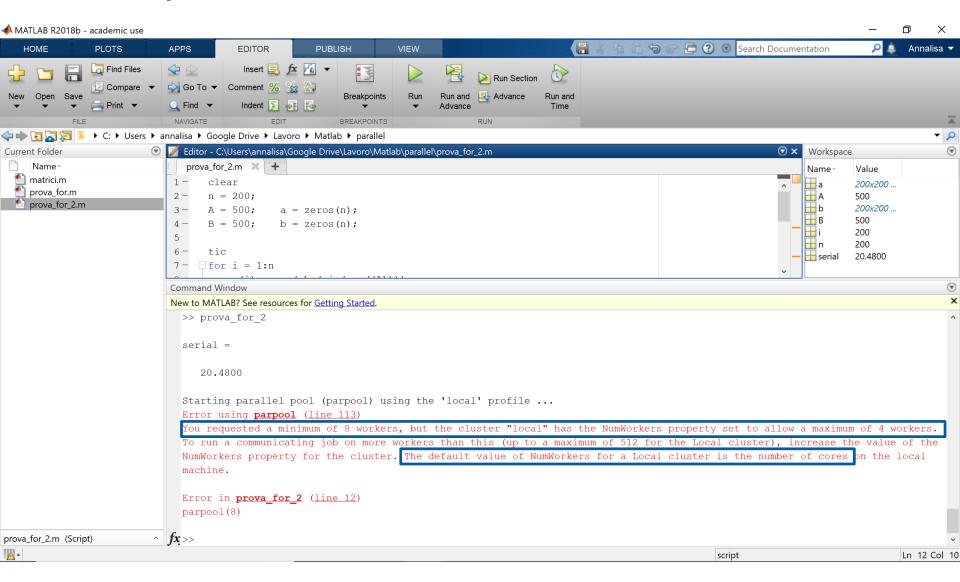


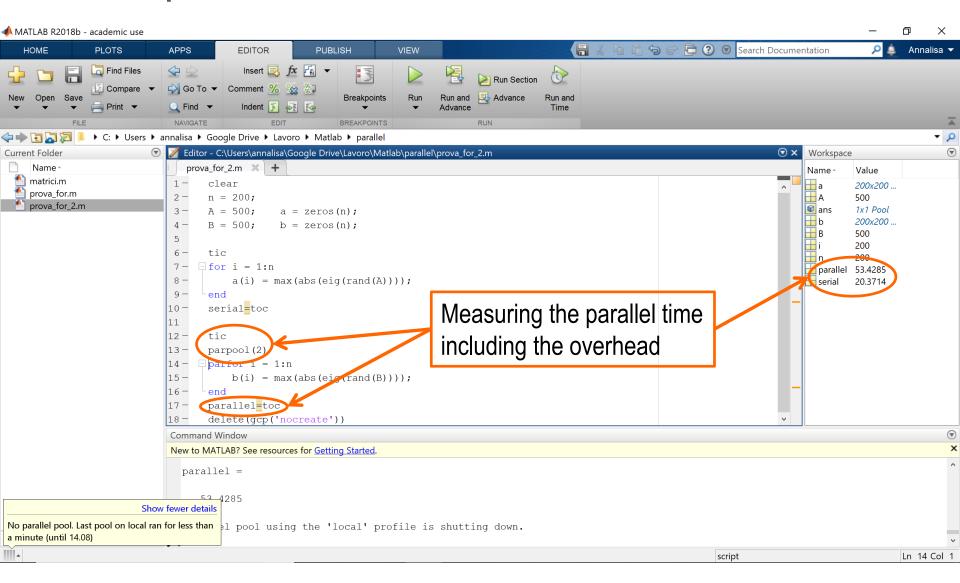


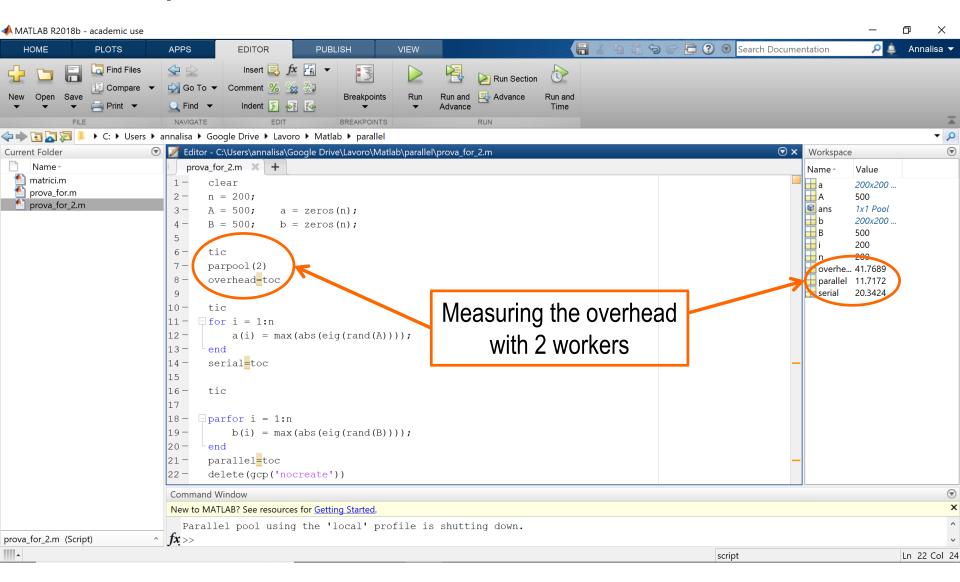


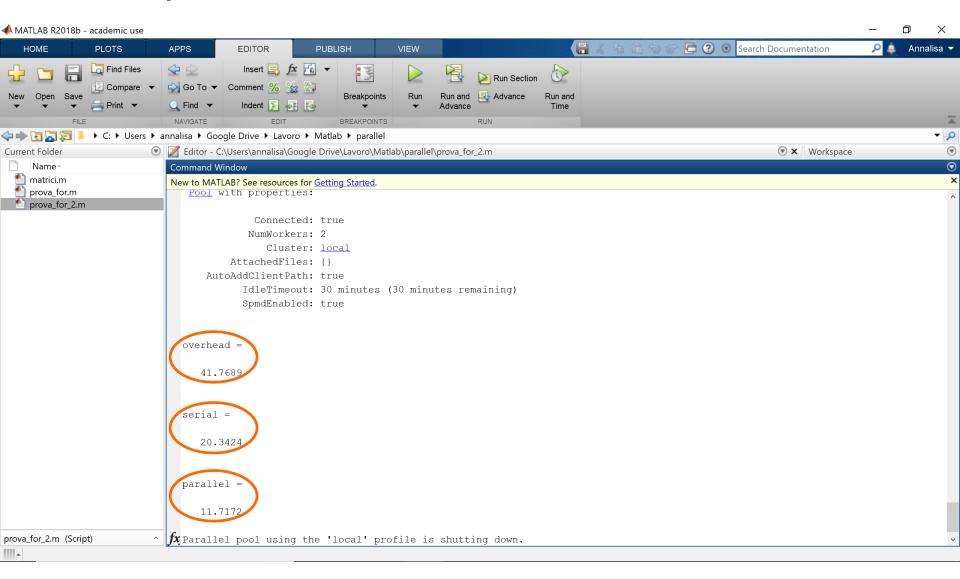


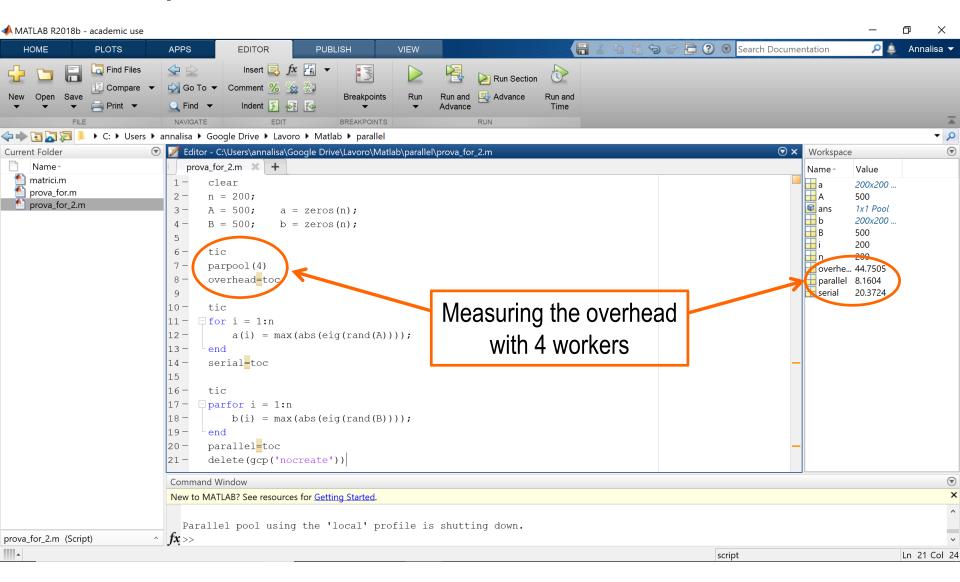












spmd

- Single Program Multiple Data model
- Used to create parallel regions of code
- Values returning from the body of an spmd statement are converted to Composite objects
- A Composite object contains references to the values stored on the remote MATLAB workers, and those values can be retrieved using cellarray indexing
- The actual data on the workers remains available on the workers for subsequent spmd execution, so long as the Composite exists on the client and the parallel pool remains open

spmd

- **spmd** distributes the array among MATLAB workers (each worker contains a part of the array) but can still operate on entire array as 1 entity
- Inside the body of the spmd statement, each MATLAB worker has:
 - a unique value of labindex,
 - the total number of workers numlabs executing the block in parallel
- Data automatically transferred between workers when necessary
- Within the body of the spmd statement, communication functions for communicating jobs (such as labSend and labReceive) can transfer data between the workers

Spmd Format

Format

```
parpool (4)
spmd
  statements
end
```

Simple Example

```
parpool(4)
spmd
  j=zeros(1e7,1);
end;
```

Spmd Examples

Result j is a Composite with 4 parts!



<1x4 Composite>

```
1: class = double, size = [10000000 1]
2: class = double, size = [10000000 1]
3: class = double, size = [10000000 1]
4: class = double, size = [10000000 1]
```

MATLAB Composites

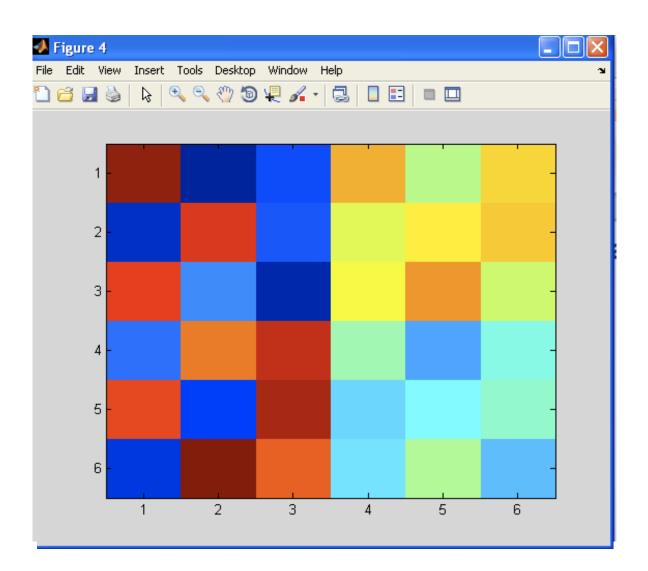
- A Composite is an object used for data distribution in MATLAB
- A Composite object has one entry for each worker
 - parpool(12) creates
 12X1 composite
 - parpool(6) creates
 6X1 composite
- You can create a composite in two ways:
 - spmd
 - c = Composite();
 - This creates a composite that does not contain any data, just placeholders for data
 - Also, one element per parpool worker is created for the composite
 - Use smpd or indexing to populate a composite created this way

Another spmd Example - creating graphs

```
%Perform a simple calculation in parallel, and plot the
 results:
 parpool(4)
  spmd
    % build magic squares in parallel
    q = magic(labindex + 2);
    % labindex - index of the lab/worker (e.g. 1)
  end
  for ii=1:length(q)
    % plot each magic square
    figure, imagesc(q{ii}); %plot a matrix as an image
  end
  delete (gcp('nocreate'));
```

Another spmd Example- creating graphs

Results



parfor vs spmd

- parfor is simpler to use
- parfor can't control iterations
- parfor only does loops
- spmd more control over iterations
- spmd more control over data movement
- spmd is persistent
- spmd is more flexible and you can create parallel regions that do more than just loop

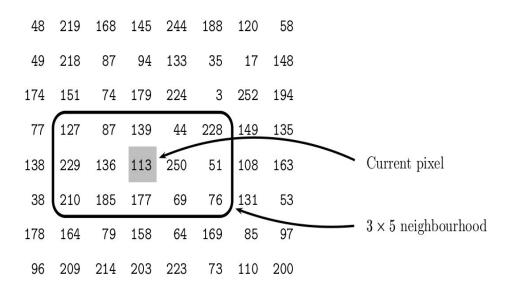
Built-in Multithreading

- Operations in the algorithm carried out by the function are easily partitioned into sections that can be executed concurrently, and with little communication or few sequential operations required
- Data size is large enough so that any advantages of concurrent execution outweigh the time required to partition the data and manage separate execution threads. For example, *most functions speed up only when the array is greater than several thousand elements*.
- Operation is not memory-bound where the processing time is dominated by memory access time. As a general rule, more complex functions speed up better than simple functions.
- http://www.mathworks.com/matlabcentral/answers/95958-which-matlab-functions-benefit-from-multithreaded-computation

IMAGES AND MATLAB

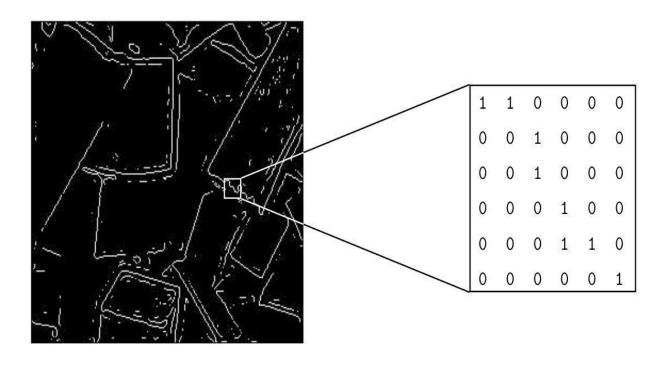
Images

- A digital image can be considered as a large array of discrete dots, each of which has a brightness associated with it
- These dots are called picture elements or more simply pixels
- The pixels surrounding a given pixel constitute its neighborhood
- A neighborhood can be characterized by its shape in the same way as a matrix: 3x3 neighborhood, 5x7 neighborhood...



Types of digital image

• Binary: Each pixel is just black or white. Since there are only two possible values for each pixel (0,1), we only need one bit per pixel



Types of digital image

- Grayscale: Each pixel is a shade of gray, normally from 0 (black) to
 255 (white), that is each pixel can be represented exactly one byte
- Other greyscale ranges can be used, generally power of 2



Types of digital image

- True Color, or RGB: Each pixel has a particular color, described by the amount of red, green and blue
- Each components has a range 0–255, for a total of 256³ different possible colors
- Three matrices representing the red, green and blue values for each pixel



64 76 82 79 78 78 93 93 91 91 86 86 88 82 88 90 88 89 125 119 113 108 111 110 137 136 132 128 126 120 105 108 114 114 118 113 96 103 112 108 111 107

Green

 66
 80
 77
 80
 87
 77

 81
 93
 96
 99
 86
 85

 83
 83
 91
 94
 92
 88

 135
 128
 126
 112
 107
 106

 141
 129
 129
 117
 115
 101

 95
 99
 109
 108
 112
 109

 84
 93
 107
 101
 105
 102

Blue

Image Import and Export

Read and write images in Matlab

```
img = imread('apple.jpg');
dim = size(img);
figure;
imshow(img);
imwrite(img, 'output.bmp', 'bmp');
```

Alternatives to imshow

```
imagesc(I)
imtool(I)
image(I)
```

Image and Matrices

How to build a matrix (or image)?

Intensity Image:

```
row = 256;
col = 256;
img = zeros(row, col);
img(100:105, :) = 0.5;
img(:, 100:105) = 1;
figure;
imshow(img);
```

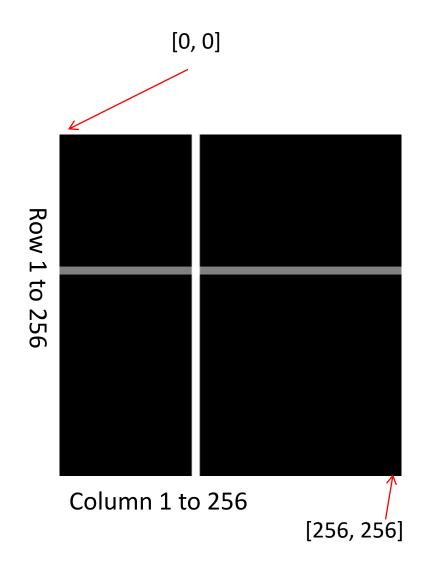


Image and Matrices

Binary Image

```
row = 256;
col = 256;
img = rand(row,
col);
img = round(img);
figure;
imshow(img);
```

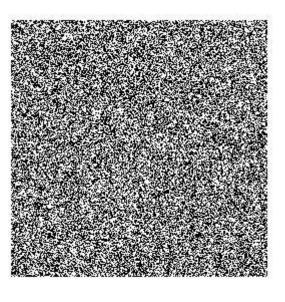


Image display

- image create and display image object
- imagesc scale and display as image
- imshow display image
- colorbar display colorbar
- getimage get image data from axes
- truesize adjust display size of image
- zoom zoom in and zoom out of 2D plot

Image information

iminfo returns information
about the image

impixel(i,j) returns the
value of the pixel(i,j)

```
Filename: 'aster.tif'
              FileModDate: '13-Mar-2008 16:54:26'
                 FileSize: 17224424.00
                   Format: 'tif'
            FormatVersion: []
                    Width: 4100.00
                   Height: 4200.00
                 BitDepth: 8.00
                ColorType: 'grayscale'
          FormatSignature: [77.00 77.00 0 42.00]
                ByteOrder: 'big-endian'
           NewSubFileType: 0
            BitsPerSample: 8.00
              Compression: 'Uncompressed'
PhotometricInterpretation: 'BlackIsZero'
             StripOffsets: [525x1 double]
          SamplesPerPixel: 1.00
             RowsPerStrip: 8.00
          StripByteCounts: [525x1 double]
              XResolution: 1.00
              YResolution: 1.00
           ResolutionUnit: 'None'
                 Colormap: []
      PlanarConfiguration: 'Chunky'
               TileWidth: []
               TileLenath: []
              TileOffsets: []
           TileByteCounts: []
              Orientation: 1.00
                FillOrder: 1.00
         GrayResponseUnit: 0.01
           MaxSampleValue: 255.00
           MinSampleValue: 0
             Thresholding: 1.00
                 Software: 'ERDAS IMAGINE '
             SampleFormat: 'Unsigned integer'
```

Image conversion

- gray2ind intensity image to index image
- im2bw image to binary
- im2double image to double precision
- im2uint8 image to 8-bit unsigned integers
- im2uint16 image to 16-bit unsigned integers
- ind2gray indexed image to intensity image
- mat2gray matrix to intensity image
- rgb2gray RGB image to grayscale
- rgb2ind RGB image to indexed image

Point Processing: Arithmetic operations

Arithmetic operations act by applying a simple function y=f(x) to each gray value in the image

- Simple functions include adding or subtract a constant value to each pixel: y = x±C (imadd, imsubtract)
- Multiplying each pixel by a constant: y = C·x (immultiply, imdivide)
- Complement: For a grayscale image is its photographic negative.

Addition





Image: I Image: I+50

Subtraction





Image: I Image: I-80

Multiplication





Image: I Image: I*3

Division





Image: I Image: I/2

Complement





Image: I Image: 255-I

- Filtering is used to enhance or attenuate some characteric of the image
- Filtering modifies the pixels in an image based on some function of a local neighborhood of each pixel



- Filtering generates a new image
- Linear filtering (cross-correlation, convolution) replace each pixel by a linear combination of its neighbors

- Linear filtering uses a matrix of coefficients W
- Imagine F is obtained from imagine I using W:

$$F[x, y] = \sum_{s=-a}^{a} \sum_{t=-b}^{b} W[s, t] I[x+s, y+t]$$

Where W and the submatrix of I are:

W[-1,-1]	W[-1,0]	W[-1,1]
W[0,-1]	W[0,0]	W[0,1]
W[1,-1]	W[1,0]	W[1,1]

I[x-1,y-1]	I[x-1,y]	I[x-1,y+1]
I[x,y-1]	I[x,y]	l[x,y+1]
I[x+1,y-1]	l[x+1,y]	I[x+1,y+1]

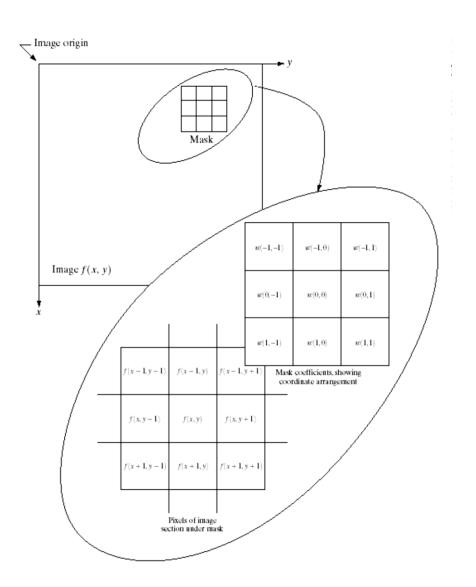
 Convolution Same as cross-correlation, except that the kernel is flipped (horizontally and vertically)

$$F[x, y] = \sum_{s=-a}^{a} \sum_{t=-b}^{b} W[s, t] I[x - s, y - t]$$

 The prescription for the linear combination - W - is called the kernel (or mask, or filter) of the cross-correlation/convolution

• Smoothing filters: mean filter, gaussian filter, median filter

Sharpening filters



Smoothing filter

Mean filter

$$W_{medio} = \frac{1}{a \cdot b} \begin{bmatrix} 1 & 1 & \cdots & 1 \\ 1 & 1 & \cdots & 1 \\ \cdots & \cdots & \cdots & \cdots \\ 1 & 1 & \cdots & 1 \end{bmatrix}$$

Smoothing filter

Gaussian filter: weights of filter follow a gaussian distribution

$$G_{\sigma}(x,y) = \frac{1}{\sigma\sqrt{2\pi}}e^{-\frac{x^2+y^2}{2\sigma^2}}$$

Example

$$G_{\sigma} = \frac{1}{273} \begin{bmatrix} 1 & 4 & 7 & 4 & 1 \\ 4 & 16 & 26 & 16 & 4 \\ 7 & 26 & 41 & 26 & 7 \\ 4 & 16 & 26 & 16 & 4 \\ 1 & 4 & 7 & 4 & 1 \end{bmatrix}$$

Gaussian filter

Removes high-frequency components from the image (low-pass filter)





Median filter

The median filter selects a sample from the window, does not average

123	125	126	130	140	
 122	124	126	127	135	
118	120	150	125	134	
 119	115	119	123	133	
111	116	110	120	130	

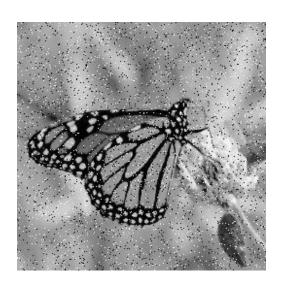
Neighbourhood values:

115, 119, 120, 123, 124, 125, 126, 127, 150

Median value: 124

Median filter

Best suited for salt and pepper noise





Sharpening filter

- Sharpening filters emphasize fine details in the image, exactly the opposite of the low-pass filter such as Gaussian filter → it just uses a different convolution kernel
- A high-pass filter can be used to make an image appear sharper.

 Usually the central pixel is positive, whereas adjacent pixels are negative

-1	-1	-1
-1	8	-1
-1	-1	-1

Sharpening filter

- First, I is modified by using a gaussian filter
- Then I_s cis obtained as a linaera combination among image I and the Gauss filtered image, with a suitable value of k usually equal to 1

$$\bar{I}[x, y] = I[x, y] - (G_{\sigma} * I)[x, y]$$
 $I_{s}[x, y] = I[x, y] + k\bar{I}[x, y]$