

# MATLAB-based Optimization

## Basic Capabilities

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3:00pm - 4:45pm, Monday, 4 February 2013  
..... FDI .....

**AOE:** Department of Aerospace and Ocean Engineering  
**ICAM:** Interdisciplinary Center for Applied Mathematics

- Introduction & *function* functions ←
- `fminbnd`
- `fminsearch`
- `lsqnonneg`
- `fzero`

# INTRO:

Basic MATLAB provides several functions for elementary minimization and zero-finding problems.

The *Optimization Toolbox* provides techniques for advanced problems, and a *gui* interface.

The OPTIMIZATION TOOLBOX is included with the VT *concurrent license*

([http://www2.ita.vt.edu/software/departement/products/mathworks/matlab\\_annual\\_concurrent/index.html](http://www2.ita.vt.edu/software/departement/products/mathworks/matlab_annual_concurrent/index.html)),

and with the VT *student license*

(<http://www2.ita.vt.edu/software/student/products/mathworks/matlab/index.html> )

but not with the VT *annual stand alone license*

([http://www2.ita.vt.edu/software/departement/products/mathworks/matlab\\_annual\\_standalone/index.html](http://www2.ita.vt.edu/software/departement/products/mathworks/matlab_annual_standalone/index.html)).

Type `ver` (enter) at the MATLAB prompt to see the list of toolboxes available with your license.

In these notes we discuss optimization and zero-finding routines in basic MATLAB. The *Optimization Toolbox* will be discussed in the next lecture.

# Basic MATLAB: linear least-squares

- Given an  $m \times n$  matrix  $A$ , a vector  $b \in \mathbb{R}^m$  with  $m > n$ .
- With more equations than unknowns we do not expect to find an  $x \in \mathbb{R}^n$  such that  $Ax = b$ .
- Instead we seek a *least squares* solution, i.e. an  $x$  that minimizes  $\|Ax - b\|^2$ .
- Such an  $x$  can be computed by `mldivide` (*matrix left-divide*)  
 $x = A \backslash b$  or  $x = \text{mldivide}(A, b)$   
If  $m = n$  and  $A$  is nonsingular,  $A \backslash b$  is similar to `inv(A)*b`, though it is not computed that way.
- Other procedures (e.g. `minres`, `gmres`) are available if  $A$  is sparse or if  $A * x$  is the output of a function, e.g. `my_Ax(x)`.
- In MATLAB `inv(A)` is rarely a good idea.

MATLAB uses this terminology to indicate that the arguments of some functions may refer to other *functions*.

A function handle is a MATLAB value that provides a means of calling a function indirectly. You can pass function handles in calls to other functions (often called *function* functions). You can also store function handles in data structures for later use. A function handle is one of the standard MATLAB data types.

```
S = functions(funhandle)
```

returns, in the MATLAB structure **S**, the function name, type, filename, and other information for the function handle stored in the variable *funhandle*. Note that `functions` is intended for debugging; do not use it your coding.

# MATLAB: *named & local functions*

A *named* function is defined in `.m` file; the name on the function line should agree with the file name (which takes precedence).

*local functions* (or *scoped functions*) are included in the same file as a named function and are visible only to it.

```
function x_out = eg_named(x_in)
% Simple example of local (sub-) functions
% We can make use of local_1 and local_2
% which are not visible to other functions
    x_out = local_1(x_in) + local_2(x_in);
%      S      = functions(@local_1)
end
% Sub functions go here
function out = local_1(in)
% Square the entries
    out = in.^2;
end
function out = local_2(in)
% Multiply by 2
    out = 2*in;
end
```

Note that each *function/subfunction* requires an `end` statement.



# MATLAB: *nested functions*

*nested functions* are included in the same file as a named function and are visible only to it. These functions are included before the `end` statement of the host function.

```
function x_out = eg_nested(x_in)
% Simple example of nested (sub-) functions
% We can make use of local_1 and local_2
% which are not visible to other functions
%
% Variables within the scope of this function are available to the nested functions

    temp = cos(x_in);
    x_out = local_1(x_in) + local_2(x_in);
%     S     = functions(@local_1)
% nested functions are 'inside' the containing functions 'end'
    function out = local_1(in)
        % Square the entries
        out = in.^2 + temp; % 'temp' is available to nested functions
    end % end for local_1
    function out = local_2(in)
        % Multiply by 2
        out = 2*in;
    end % end for local_2

end % end for eg_nested
```

# MATLAB: a simple named function - my\_func

A simple named function

```
function y = my_func(x, param)
%Simple parameterized scalar function
    y = -x.*cos(param*x);
end
```

Examine with functions()

```
>> S = functions(@my_func)
S =
    function: 'my_func'
           type: 'simple'
           file: [1x70 char]
```

```
>> S.file
```

```
ans =
/Users/ecliff/Documents/classes/fdi_matlab_2013/optimization/sample_code/my_func.m
```



*anonymous functions* are defined locally and do not exist in a file.

To assign a handle to an anonymous function use the '@' symbol

```
% This can be a function or a script
%
% Variables defined in this code can be used
% as arguments in defining an anonymous function

parameter = stuff;

%% Define the (handle to the) anonymous function
fun_hdl = @(x) my_func(x, parameter);

%% Use the anonymous function
temp = feval(fun_hdl, 3.2);
```

`feval` is a *built-in function* whose first argument is a *function*

## MATLAB: using functions() again

```
param = 2;
fun_hdl = @(x) my_func(x, param);
param = 4;
S = functions(fun_hdl)
S =
    function: '@(x)my_func(x,param)'
           type: 'anonymous'
           file: ''
    workspace: {[1x1 struct]}
fields(S.workspace{1})
ans =
    'param'
S.workspace{1}.param
ans =
    2
```

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- `lsqnonneg`
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## Basic MATLAB: fminbnd

fminbnd computes the minimum of a scalar-valued function of a scalar variable within a given interval .

```
>> param = 2;
>> fun_hndl = @(x) my_func(x, param);
>> opt = optimset('fminbnd');
>> opt = optimset(opt, 'Display', 'iter');
>> xs = fminbnd(fun_hndl, 0, 1, opt);
```

Func-count	x	f(x)	Procedure
1	0.381966	-0.275826	initial
2	0.618034	-0.203032	golden
.... 3 lines omitted			
6	0.430208	-0.280548	parabolic
7	0.430161	-0.280548	parabolic
8	0.430127	-0.280548	parabolic

Optimization terminated:

the current x satisfies the termination  
criteria using OPTIONS.TolX of 1.000000e-04



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fminbnd can be invoked with additional output arguments

doc fminbnd

for details

## Syntax

```
x = fminbnd(fun,x1,x2)
```

```
x = fminbnd(fun,x1,x2,options)
```

```
[x,fval] = fminbnd(...)
```

```
[x,fval,exitflag] = fminbnd(...)
```

```
[x,fval,exitflag,output] = fminbnd(...)
```

- Introduction & *function* functions
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- `fminsearch` computes a candidate for the minimum of a scalar-valued function of a vector variable from a given starting point .
- With  $x \in \mathbb{R}^n$  the procedure is based on a *simplex* of  $(n + 1)$  points, and proceeds to produce a test point to replace the *worst* point in the simplex.
- The MATLAB algorithm begins by evaluating the function at  $x_0$  and the  $n$  perturbed points  $x_i = x_0 + \delta * e_i$ , with  $\delta = 0.05\|x_0\|$ . The points are re-labeled to  $x_j$ ,  $j = 1, \dots, n + 1$  so that  $f(x_j) \leq f(x_{j+1})$ , i.e. in ascending order.
- Compute  $\bar{x}$ , the average over the set  $\{x_k \mid k = 1, \dots, n\}$ .

The algorithm then proceeds as described in the following figures:

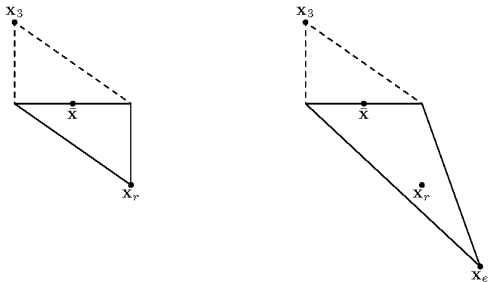


FIG. 1. Nelder-Mead simplices after a reflection and an expansion step. The original simplex is shown with a dashed line.



or:

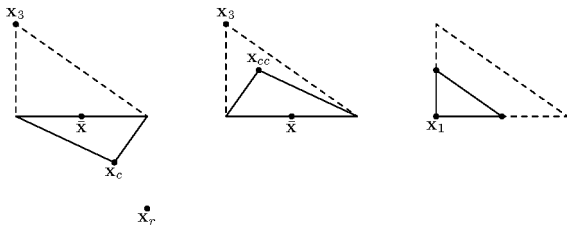


FIG. 2. Nelder-Mead simplices after an outside contraction, an inside contraction, and a shrink. The original simplex is shown with a dashed line.

# fminsearch for Rosenbrock's example

```
% Script to display use of fminsearch on the Rosenbrock valley function

%
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%
% FDI Short course on Matlab-based Optimization – basic capabilities

% Example using the Rosenbrock valley function
% The first term in the sum is zero on the parabola  $x(2)=x(1).^2$ , and
% increases rapidly as  $x(:)$  departs from the 'valley'.

f_hndl = @(x) 100*(x(2) - x(1)*x(1))^2 + (1 - x(1))^2; % anonymous function

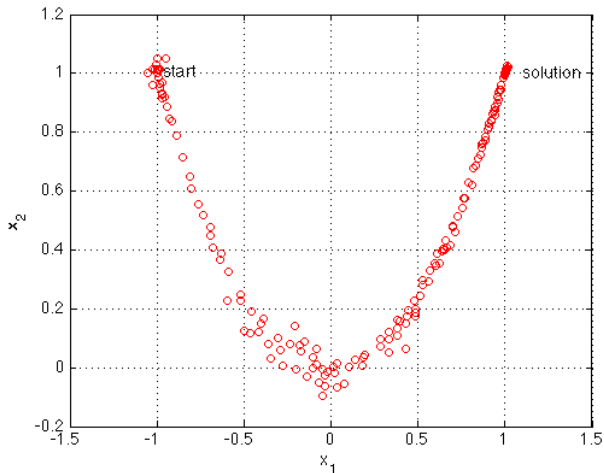
x0      = [-1; 1]; % starting point

opt     = optimset('fminsearch');
opt     = optimset(opt, 'Display', 'iter');

xs = fminsearch(f_hndl, x0, opt); % invoke Nelder Meade
```

Iteration	Func-count	min f(x)	Procedure
0	1	4	
1	3	4	initial simplex
2	5	4	contract inside
3	7	4	contract inside
4	9	4	contract outside
5	11	4	contract inside
6	13	3.93698	expand
99	185	2.64787e-09	contract outside
100	187	4.05759e-10	contract inside

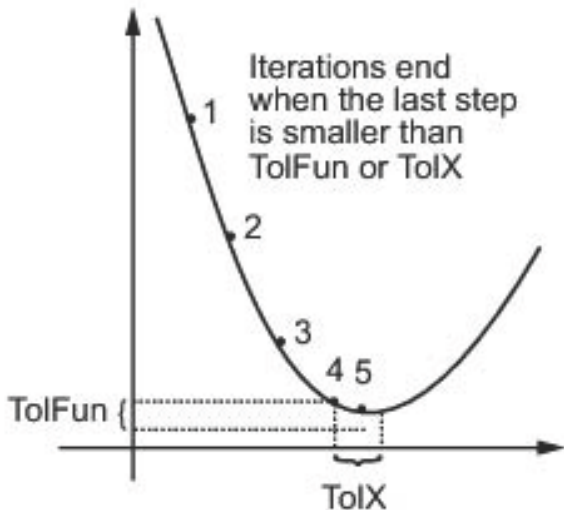
# fminsearch for Rosenbrock's example



**Syntax** options =  
optimset('param1',value1,'param2',value2,...)  
options = optimset(optimfun)  
options = optimset(oldopts,'param1',value1,...)  
options = optimset(oldopts,newopts)

Which parameters are meaningful depends on the function and the *algorithm*.

# Interpreting TolFun and TolX

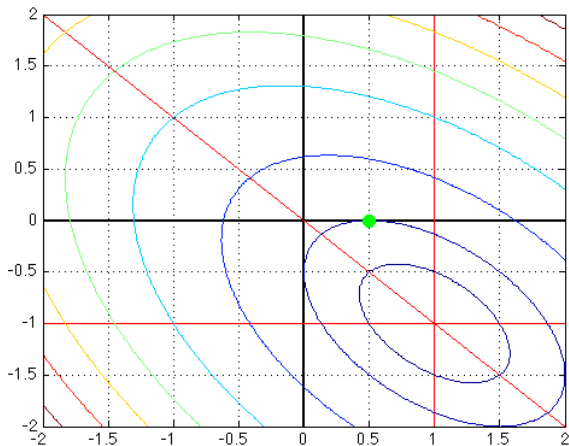


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```
x = lsqnonneg( A, b)  
minimize  $\|Ax - b\|^2$  subject to  $x \geq 0$ 
```

```
>> A = [ 1 1; 1 0; 0 1];  
>> b = [0; 1; -1];  
>> x1 = lsqnonneg(A, b)
```

```
x1 =  
  
    0.5000  
         0
```





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- `xs = fzero(fun, x0)` tries to find a zero of the scalar-valued function `fun` near the scalar `x0`.
- if `x0` is a vector of length 2, then `fzero` assumes that `fun(x0(1))` and `fun(x0(2))` have opposite signs. Assuming that `fun` is continuous, `fzero` will return a scalar near where `fun` changes sign.
- If unsuccessful, `fzero` will return a NaN.
- `fzero(fun, x0, optim)` allows one to set `OPTIONS`.

## fzero: an example

```
>> fh = @(x) sin(x)
```

```
fh =
```

```
    @(x)sin(x)
```

```
>> fzero(fh, 1)
```

```
ans =
```

```
    1.5485e-24
```

```
>> fzero(fh, 3)
```

```
ans =
```

```
    3.1416
```

- In next week's class we'll cover the  
OPTIMIZATION TOOLBOX  
particularly `fmincon`.
- We shall use today's material on nested functions

Please complete the evaluation form  
<http://www.fdi.vt.edu/training/evals/>

Thanks