

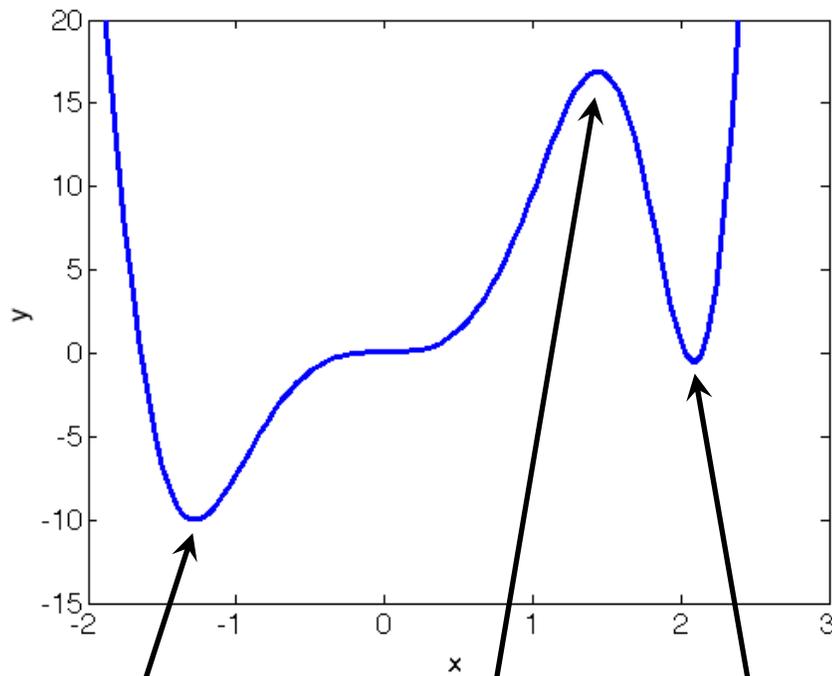


## Week 5 Optimization

### 1-D example

$$f(x) = x^4 + 10x \sin x^2$$

```
f = @(x) x.^4 + 10*x.*sin(x.^2);  
x = linspace(-2,3,100);  
plot(x,f(x));
```



```
[xmin, fmin]=fminbnd(f, -2, -1);  
[xmin, fmin]=fminsearch(f, -1);  
[xmin, fmin]=fminunc(f, -1);
```

```
[xmin, fmin]=fminbnd(f, 1.5, 2.5);  
[xmin, fmin]=fminsearch(f, 2);  
[xmin, fmin]=fminunc(f, 2);
```

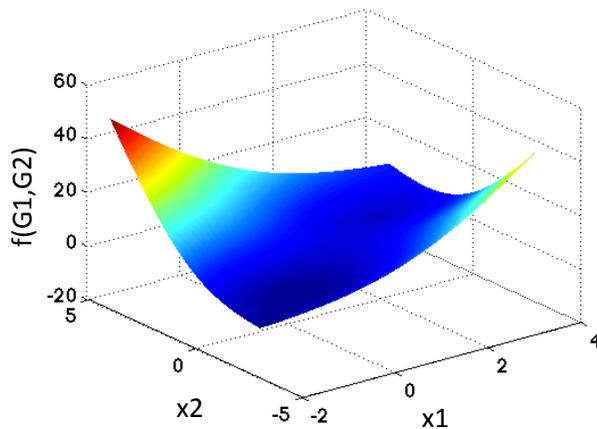
```
[xmax, fmax]=fminbnd(@(x) -f(x), 1, 2); fmax=-fmax;  
[xmax, fmax]=fminsearch(@(x) -f(x), 1); fmax=-fmax;  
[xmax, fmax]=fminunc(@(x) -f(x), 1); fmax=-fmax;
```

## 2-D example

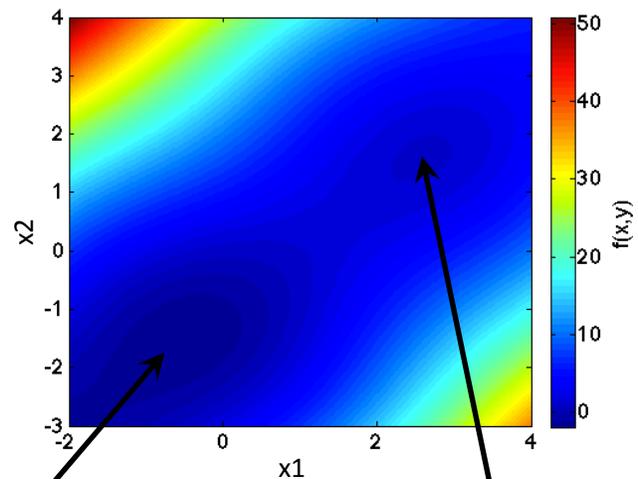
$$f(x_1, x_2) = \sin(x_1 + x_2) + (x_1 - x_2)^2 - 1.5x_1 + 2.5x_2 + 1$$

```
f = @(x1,x2) sin(x1 + x2) + (x1 - x2).^2 - 1.5*x1 + 2.5*x2 + 1;
x1 = -2:0.05:4;
x2 = -3:0.05:4;
[G1,G2] = meshgrid(x1,x2);
```

```
surf(x1,x2,f(G1,G2));
```



```
contourf(x1,x2,f(G1,G2));
```



```
g = @(x) sin(x(1) + x(2)) + (x(1) - x(2)).^2 - 1.5*x(1) + 2.5*x(2) + 1;
[xmin,fmin]=fminunc(g, [-1,-2]);
[xmin,fmin]=fminsearch(g, [-1,-2]);
>> xmin = -0.5472    -1.5472
>> fmin = -1.9132
```

```
[xmin,fmin]=fminunc(g, [3,2]);
[xmin,fmin]=fminsearch(g, [3,2]);
>> xmin = 2.5944    1.5944
>> fmin = 1.2284
```

Note: `fminsearch` and `fminbnd` only accept function handles with one input argument. So instead of `f`, we have defined a new function handle `g` for the use in `fminsearch` and `fminbnd`. Alternatively, if you want to recycle `f`, you can do the following:

```
[xmin,fmin]=fminbnd(@(x) f(x(1),x(2)), [-1,-2]);
[xmin,fmin]=fminsearch(@(x) f(x(1),x(2)), [-1,-2]);
>> xmin = -0.5472    -1.5472
>> fmin = -1.9132
```

Algorithms:

	Derivative	Derivative-free
1-D	<p>Newton <code>fminunc (f, x0)</code></p> <p>local quadratic approximation to f</p> <p><math>x_0</math>, <math>x_1</math>, <math>x_2</math></p> <p><math>f(x)</math></p>	<p>Golden Section Search (<math>\varphi = 0.61834</math>) Successive Parabolic <code>fminbnd (f, x1, x2)</code></p> <p><math>x_1</math>, <math>x_3</math>, <math>x_4</math>, <math>x_2</math></p> <p><math>f(x)</math></p> <p><math>k=1</math></p> <p>0, -4, -8, -12</p> <p>-1.5, -1, -0.5</p>
N-D	<p>Steepest Descent Conjugate Gradient <code>fminunc (f, x0)</code></p> <p><math>\nabla F</math></p>	<p>Nelder-Mead <code>fminsearch (f, x0)</code></p> <p>2: Reflection</p> <p>3: Expansion</p> <p>1: Matlab</p> <p>4</p> <p>5: Reduction</p> <p>4: Contraction</p> <p><math>x_3</math>, <math>x_5</math>, <math>x_{min}</math>, <math>x_2</math>, <math>x_4</math>, <math>x_0</math>, <math>x_1</math></p> <p>User-supplied guess</p> <p>Matlab's choice = <math>1.05 x_0</math></p>

## Linear Programming

Simplex methods: **linprog** — Check vertices (because of linearity)

**linprog(p,A,b,Aeq,beq,lb,ub,options)**

$$\text{where } f(\mathbf{x}) = \mathbf{p}^T \mathbf{x} \text{ subject to } \begin{cases} \mathbf{Ax} \leq \mathbf{b} \\ \mathbf{A}_{eq} \mathbf{x} = \mathbf{b}_{eq} \\ \mathbf{l}_{eq} \leq \mathbf{x} \leq \mathbf{u}_{eq} \end{cases}$$

Example:

$$f(x_1, x_2) = 2x_1 + 7x_2 \text{ subject to } \begin{cases} x_1 + 2x_2 \geq 3 \\ 3x_1 + x_2 \leq 21 \\ x_1, x_2 \geq 0 \\ x_2 \leq 10 \end{cases} \Rightarrow \begin{cases} -x_1 - 2x_2 \leq -3 \\ 3x_1 + x_2 \leq 21 \\ 0 \leq x_1 \leq \infty \\ 0 \leq x_2 \leq 10 \end{cases} \Rightarrow \begin{cases} \begin{bmatrix} -1 & -2 \\ 3 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \leq \begin{bmatrix} -3 \\ 21 \end{bmatrix} \\ \begin{bmatrix} 0 \\ 0 \end{bmatrix} \leq \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \leq \begin{bmatrix} \infty \\ 10 \end{bmatrix} \end{cases}$$

`A = [-1 -2; 3 1];`

`b = [-3; 21];`

`lb = [0; 0];`

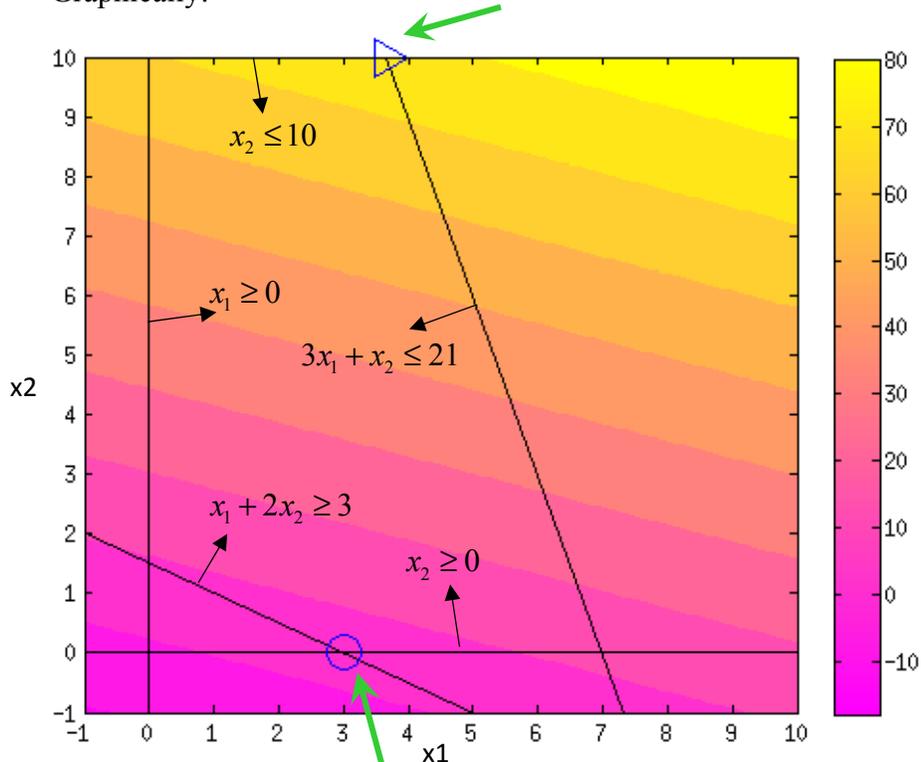
`ub = [inf; 10];`

`p = [2; 7];`

`[xmin,fmin] = linprog(p,A,b,[],[],lb,ub);`

`[xmax,fmax] = linprog(-p,A,b,[],[],lb,ub); fmax=-fmax;`

Graphically: `>> xmax = 3.6667 10.0000` `>> fmax = 77.3333`



`>> xmin = 3.0000 0.0000`

`>> fmin = 6.0000`