## Example 1

consider the function: $f(x, y)=x^{2}-y^{2}$

$$
\frac{\partial f}{\partial x}=2 x \text { and } \frac{\partial f}{\partial y}=-2 y
$$

These first derivatives are zero at $\mathrm{x}^{*}=0$ and $\mathrm{y}^{*}=0$. The Hessian matrix of $f$ is:

$$
\mathrm{H}(f)=\left[\begin{array}{cc}
\frac{\partial^{2} f}{\partial x^{2}} & \frac{\partial^{2} f}{\partial x \partial y} \\
\frac{\partial^{2} f}{\partial y \partial x} & \frac{\partial^{2} f}{\partial y^{2}}
\end{array}\right]=\left[\begin{array}{l}
2 \\
0
\end{array}\right] \begin{gathered}
0 \\
-2
\end{gathered}
$$

The Hessian matrix of $f$ at $\left(\mathrm{x}^{*}, \mathrm{y}^{*}\right)=\mathrm{H}$
The determinant $\mathrm{H} 1=2$ (positive), and the determinant $\mathrm{H} 2=2(-2)-0(0)=-4$ (negative). Then H is indefinite.

Since this matrix is neither positive definite nor negative definite, the point ( $\mathrm{x}^{*}=0, \mathrm{y}^{*}=0$ ) is a saddle point.

## Example 2

## Find the critical points of the function:

$$
f\left(x_{1}, x_{2}\right)=x_{1}^{3}+x_{2}^{3}+2 x_{1}^{2}+4 x_{2}^{2}+6
$$

SOLUTION: The necessary conditions for the existence of an extreme point a $\frac{\partial f}{\partial x_{1}}=3 x_{1}^{2}+4 x_{1}=x_{1}\left(3 x_{1}+4\right)=0$

$$
\frac{\partial f}{\partial x_{2}}=3 x_{2}^{2}+8 x_{2}=x_{2}\left(3 x_{2}+8\right)=0
$$

From (1) $x_{1}=0$ or $\left(-\frac{4}{3}\right)$, and from (2) $x_{2}=0$ or $\left(-\frac{8}{3}\right)$. Then these equations are satisfied at the points:
$(0,0),\left(0,-\frac{8}{3}\right),\left(-\frac{4}{3}, 0\right)$, and $\left(-\frac{4}{3},-\frac{8}{3}\right)$

To find the nature of these extreme points, we have to use the sufficiency conditions. The secondorder partial derivatives of $\boldsymbol{f}$ are given by:

$$
\begin{array}{ll}
\frac{\partial f}{\partial x_{1}}=3 x_{1}^{2}+4 x_{1} & \frac{\partial f}{\partial x_{2}}=3 x_{2}^{2}+8 x_{2} \\
\frac{\partial^{2} f}{\partial x_{1}^{2}}=6 x_{1}+4 & \frac{\partial^{2} f}{\partial x_{2}^{2}}=6 x_{2}+8
\end{array} \frac{\partial^{2} f}{\partial x_{1} \partial x_{2}}=0
$$

The Hessian matrix of $f$ is given by:

$$
\mathrm{H}\left(f\left[\begin{array}{cc}
\frac{\partial^{2} f}{\partial x_{1}^{2}} & \frac{\partial^{2} f}{\partial x_{1} \partial x_{2}} \\
\frac{\partial^{2} f}{\partial x_{2} \partial x_{1}} & \frac{\partial^{2} f}{\partial x_{2}^{2}}
\end{array}\right]=\left[\begin{array}{cc}
6 x_{1}+4 & 0 \\
0 & 6 x_{2}+8
\end{array}\right]\right.
$$

$$
f\left(x_{1}, x_{2}\right)=x_{1}^{3}+x_{2}^{3}+2 x_{1}^{2}+4 x_{2}^{2}+6 \quad \mathbf{H}(f)=\left[\begin{array}{cc}
6 x_{1}+4 & 0 \\
0 & 6 x_{2}+8
\end{array}\right]
$$

$$
\text { If } J_{1}=\left|6 x_{1}+4\right| \text { and } J_{2}=\left|\begin{array}{cc}
6 x_{1}+4 & 0 \\
0 & 6 x_{2}+8
\end{array}\right| \text {, the values of } J_{1} \text { and } J_{2} \text { and }
$$

the nature of the extreme point are as given below.

|  | Value <br> of $J_{1}$ | Value <br> of $J_{2}$ | Nature of $\mathbf{J}$ | Nature of $\mathbf{X}$ | $f(\mathbf{X})$ |
| :--- | :---: | :---: | :--- | :--- | :---: | :---: |
| Point $\mathbf{X}$ | +4 | +32 | Positive definite | Relative minimum | 6 |
| $(0,0)$ | +4 | -32 | Indefinite | Saddle point | $418 / 27$ |
| $\left(0,-\frac{8}{3}\right)$ | -4 | -32 | Indefinite | Saddle point | $194 / 27$ |
| $\left(-\frac{4}{3}, 0\right)$ | -4 | +32 | Negative definite | Relative maximum | $50 / 3$ |
| $\left(-\frac{4}{3},-\frac{8}{3}\right)$ | -4 |  |  |  |  |

### 1.7 CONSTRAINED PROBLEMS

### 1.7.1 Multivariable Optimization With Equality Constraints

We consider the optimization of continuous functions subjected to equality constraints:

$$
\begin{aligned}
& \text { Minimize } f=f(\mathbf{X}) \\
& \text { subject to } \\
& g_{j}(\mathbf{X})=0, \quad j=1,2, \ldots, m
\end{aligned}
$$

$$
\text { Where: } \mathbf{x}=\left\{\begin{array}{c}
x_{1} \\
x_{2} \\
\vdots \\
x_{n}
\end{array}\right\}
$$

Here $m$ is less than or equal to $n$; otherwise (if $m>n$ ), the problem becomes overdefined and, in general, there will be no solution. There are several methods available for the solution of this problem: The Constrained variation, Jacobian method, Methods of direct substitution, and Lagrange multipliers.

### 1.7.1.1 Method of Direct Substitution

For a problem with $n$ variables and $m$ equality constraints, it is theoretically possible to solve simultaneously the $m$ equality constraints and express any set of $m$ variables in terms of the remaining $n-m$ variables. When these expressions are substituted into the original objective function, there results a new objective function involving only $n-m$ variables. The new objective function is not subjected to any constraint, and hence its optimum can be found by using the unconstrained optimization techniques.

## Example 1

Minimize: $f(\mathbf{x})=4 x_{1}^{2}+5 x_{2}^{2}$
Subject to: $2 x_{1}+3 x_{2}=6$
Either $\mathrm{x}_{1}$ or $\mathrm{x}_{2}$ can be eliminated without difficulty. Solving for $\mathrm{x}_{1}$,

$$
x_{1}=\frac{6-3 x_{2}}{2}
$$

Substitute for $x_{1}$ in the Objective Function, the new equivalent objective function in terms of a single variable $x_{2}$ is:
$f\left(x_{2}\right)=14 x_{2}^{2}-36 x_{2}+36$
The constraint in the original problem has now been eliminated, and $f\left(x_{2}\right)$ is an unconstrained function with one independent variable.

We can now minimize the new objective function by setting the first derivative of $f$ equal to zero, and solving for the optimal value of $x_{2}$ :

$$
\frac{d f\left(x_{2}\right)}{d x_{2}}=28 x_{2}-36=0 \quad x_{2}^{*}=1.286
$$

$$
f\left(x_{2}\right)=14 x_{2}^{2}-36 x_{2}+36
$$

$f^{\prime \prime}(\mathrm{x})=28$ (positive), then X * is a local minimum.
Once $\mathrm{x}_{2}{ }^{*}$ is obtained, then, $\mathrm{x}_{1}{ }^{*}$ can be directly obtained via the relation (1): $x_{1}=\frac{6-3 x_{2}}{2}$, then: $x_{1}^{*}=\frac{6-3 x_{2}^{*}}{2}=1.071$

$$
\begin{aligned}
& f(\mathbf{x})=4 x_{1}^{2}+5 x_{2}^{2} \\
& f_{\text {min }}=4(1.071)^{2}+5(1.286)^{2}=12.85714
\end{aligned}
$$

## Solving Using Excel

## Problem: Minimize: $f(\mathbf{x})=4 x_{1}^{2}+5 x_{2}^{2}$ <br> Subject to: $2 x_{1}+3 x_{2}=6$

| F4 |  | $\checkmark$ | $f_{x}$ | =SUMPRODUCT(B4:C4;B3:C3) |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 4 | A | B | C | D | E | F | G | H | 1 |
| 1 |  |  |  |  |  |  |  |  |  |
| 2 | terms | x1^2 | x2^2 | x1 | x2 |  |  |  |  |
| 3 | value | 0 | 0 |  |  |  |  |  |  |
| 4 | f | 4 | 5 |  |  | 0 |  |  |  |
| 5 |  |  |  |  |  |  |  |  |  |
| 6 | constraint |  |  | 2 | 3 | 0 | = | 6 |  |
| 7 |  |  |  |  |  |  |  |  |  |

$$
\begin{aligned}
& \mathrm{B} 3:=\mathrm{D} 3^{\wedge} 2 \\
& \mathrm{C} 3:=\mathrm{E} 3^{\wedge} 2 \\
& \mathrm{~F} 4:=\mathrm{SUMPRODUCT}(\mathrm{~B} 4: \mathrm{C} 4 ; \mathrm{B} 3: \mathrm{C} 3) \\
& \mathrm{F} 6:=\text { SUMPRODUCT(D6:E6;D3:E3) }
\end{aligned}
$$







## Example 2

The profit analysis model:
Max the profit $z=v . p-c_{f}-v . c_{v}$
The demand is represented by: $v=1,500-\mathbf{2 4 . 6 p}$
Where: $v=$ volume (quantity), $p=$ price,

$$
\begin{equation*}
c_{f}=\text { fixed cost }=\$ 10,000, c_{v}=\text { variable cost }=\$ 8 \text { per unit. } \tag{2}
\end{equation*}
$$

Substituting values of $c_{f}$ and $c_{v}$ into (1), we obtain:
$\mathrm{z}=\mathrm{v} . \mathrm{p}-10,000-\mathbf{8 v}$
Substituting (2) in (3):
$\left.\mathrm{z}=1500 \mathrm{p}-24.6 \mathrm{p}^{2}-\mathbf{1 0 , 0 0 0}-\mathbf{8 ( 1 , 5 0 0 - 2 4 . 6 p}\right)$
$\mathrm{z}=1696.8 \mathrm{p}-24.6 \mathrm{p}^{2}-22,000$
$\frac{d z}{d p}=1696.8 \mathrm{p}-49.2 \mathrm{p}=0$ for the critical points, then:
$\mathrm{p}^{*}=34.49$
$\frac{d^{2} z}{d p^{2}}=-49.2$ (negative), then $\mathbf{p}^{*}$ is a local maximum.
Substituting in (2): $v^{*}=1500-24.6(34.49)=651.55$
Substituting in (3): $\mathrm{z}_{\text {max }}=(651.55)(34.49)-10,000-8(651.55)=7259.56$

### 1.7.1.2 Lagrange Method

The basic features of the Lagrange multiplier method is given initially for a simple problem of two variables with one constraint.
The extension of the method to a general problem of $n$ variables with $m$ constraints is given later.

## Problem with Two Variables and One Constraint.

Consider the problem Minimize $f\left(x_{1}, x_{2}\right)$
subject to: $g\left(x_{1}, x_{2}\right)=0$
Define Lagrange function $L\left(x_{1}, x_{2}, \lambda\right)=f\left(x_{1}, x_{2}\right)+\lambda g\left(x_{1}, x_{2}\right)$
$\lambda$ is called the Lagrange multiplier.
$L$ is treated as a function of the three variables $\mathrm{x}_{1}, \mathrm{x}_{2}$, and $\lambda$.
Theorem: Necessary Conditions for Extremum:

$$
\begin{aligned}
& \frac{\partial L}{\partial x_{1}}\left(x_{1}, x_{2}, \lambda\right)=\frac{\partial f}{\partial x_{1}}\left(x_{1}, x_{2}\right)+\lambda \frac{\partial g}{\partial x_{1}}\left(x_{1}, x_{2}\right)=0 \\
& \frac{\partial L}{\partial x_{2}}\left(x_{1}, x_{2}, \lambda\right)=\frac{\partial f}{\partial x_{2}}\left(x_{1}, x_{2}\right)+\lambda \frac{\partial g}{\partial x_{2}}\left(x_{1}, x_{2}\right)=0 \\
& \frac{\partial L}{\partial \lambda}\left(x_{1}, x_{2}, \lambda\right)=g\left(x_{1}, x_{2}\right)=0
\end{aligned}
$$

## Theorem: Sufficient Condition

A sufficient condition for $f(X)$ to have a relative minimum at $X^{*}$ is that the quadratic, $Q$, defined by:

$$
\mathrm{Q}=\frac{\partial^{2} L}{\partial x_{1} \partial x_{2}} d x_{1} d x_{2}
$$

evaluated at $\mathrm{X}=\mathrm{X}$ * must be positive definite for all values of $d X$ for which the constraints are satisfied.

- If Q is negative definite for all choices of the admissible variations $d X, \mathrm{X}^{*}$ will be a constrained maximum of $f(\mathrm{X})$.
- It has been shown by Hancock that a necessary condition for the quadratic form $Q$, to be positive (negative) definite for all admissible variations $d \mathrm{X}$ is that each root of the polynomial $Z$, defined by the following determinantal equation, be positive (negative):

$$
\left|\begin{array}{ccc}
L_{11}-z & L_{12} & g_{11} \\
L_{21} & L_{22}-z & g_{12} \\
g_{11} & g_{12} & 0
\end{array}\right|=0
$$

where

$$
\begin{aligned}
& L_{11}=\left.\frac{\partial^{2} L}{\partial x_{1}^{2}}\right|_{\left(\mathbf{N}^{*}, \lambda^{*}\right)} L_{12}=\left.\frac{\partial^{2} L}{\partial x_{1} \partial x_{2}}\right|_{\left.\mathbf{(}^{*}, \lambda^{*}\right)}=L_{21} \quad L_{22}=\left.\frac{\partial^{2} L}{\partial x_{2}^{2}}\right|_{\left(\mathbf{X}^{*}, \lambda^{*}\right)}=0 \\
& g_{11}=\left.\frac{\partial g_{1}}{\partial x_{\mathbf{1}}}\right|_{\left(\mathbf{X}^{*}, \lambda^{*}\right)} \quad g_{12}=\left.\frac{\partial g_{1}}{\partial x_{2}}\right|_{\left(\mathbf{X}^{*}, \lambda^{*}\right)}
\end{aligned}
$$

## Example

Find the solution of the following problem using the Lagrange multiplier method:
$f(x, y)=x^{-1} y^{-2}$
Subject to: $g(x, y)=x^{2}+y^{2}-4=0$
The Lagrange function is:
$L(x, y, \lambda)=f(x, y)+\lambda g(x, y)=x^{-1} y^{-2}+\lambda\left(x^{2}+y^{2}-4\right)$
The necessary conditions for the extreme of $f(x, y)$ give:
$\frac{\partial L}{\partial x}=-x^{-2} y^{-2}+2 \lambda x=0$
(1) $\Rightarrow \lambda=\frac{1}{2} x^{-3} y^{-2} \ldots \ldots \ldots$ (4)
$\frac{\partial L}{\partial y}=-2 x^{-1} y^{-3}+2 \lambda y=0$
(2) $\Rightarrow \lambda=x^{-1} y^{-4}$
$\frac{\partial L}{\partial \lambda}=x^{2}+y^{2}-4=0$
(3) From (4), (5): $\frac{1}{2} x^{-3} y^{-2}=x^{-1} y^{-4}$
$\frac{1}{2} x^{-3} y^{-2}=x^{-1} y^{-4} \Rightarrow \frac{1}{2} y^{-2}=x^{2} y^{-4} \Rightarrow \frac{1}{2} y^{2}=x^{2}$
$X^{*}=\frac{1}{\sqrt{2}} y^{*} \quad$ Or: $\quad x^{2}=\frac{1}{2} y^{2}$
(3): $x^{2}+y^{2}-4=0$

From (6): $\frac{1}{2} y^{2}+y^{2}=4 \Rightarrow y^{2}=\frac{8}{3} \Rightarrow$
$\mathbf{y}^{*}=2 \sqrt{\frac{2}{3}}$
Substituting in (6): $x^{*}=\frac{2}{\sqrt{3}}$
(5): $\lambda=x^{-1} \boldsymbol{y}^{-4} \square \quad \lambda *=\frac{\sqrt{3}}{2} \frac{1}{16} \frac{\sqrt{81}}{\sqrt{16}}=\frac{9 \sqrt{3}}{128}$

The determinant:
$\left|\begin{array}{ccc}L_{11}-z & L_{12} & g_{11} \\ L_{21} & L_{22}-z & g_{12} \\ g_{11} & g_{12} & 0\end{array}\right|=0$
$\left|\frac{\partial^{2} L}{\partial x^{2}}\right|_{\left(\mathbf{X}^{*}, \lambda^{*}\right)}-\mathbf{z}$
$\left.\left.\frac{\partial^{2} L}{\partial x_{1} \partial x_{2}}\right|_{\left(\mathbf{X}^{*}, \lambda^{*}\right)} \quad \frac{\partial g_{1}}{\partial x_{1}}\right|_{\left(\mathbf{X}^{*}, \lambda^{*}\right)}$
$\left|\begin{array}{ccc}\left.\frac{\partial^{2} L}{\partial x_{1} \partial x_{2}}\right|_{\left(\mathbf{N}^{*}, \lambda^{*}\right)} & \left.\frac{\partial^{2} L}{\partial x_{2}^{2}}\right|_{\left(\mathbf{X}^{*}, \lambda^{*}\right)} & -\mathbf{z} \\ \left.\frac{\partial g_{1}}{\partial x_{1}}\right|_{\left(\mathbf{X}^{*}, \lambda^{*}\right)} & \left.\frac{\partial g_{1}}{\partial x_{2}}\right|_{\left(\mathbf{X}^{*}, \lambda^{*}\right)} \\ & x_{2} & \left.\right|_{\left.\mathbf{X}^{*}, \lambda^{*}\right)} \\ 0\end{array}\right|=0$
$\left|\begin{array}{ccc}1.218-\mathrm{z} & 0.344 & 2.309 \\ 0.344 & 0.974-\mathrm{z} & 3.266 \\ 2.309 & 3.266 & 0\end{array}\right|=0$
$\left|\begin{array}{lcc}1.218-\mathrm{z} & 0.344 & 2.309 \\ 0.344 & \begin{array}{cc}0.974-\mathrm{z} & 3.266 \\ 2.309 & 3.266\end{array} & 0\end{array}\right|=$
$(1.218-\mathrm{z})\left|\begin{array}{cc}0.974-\mathrm{z} & 3.266 \\ 3.266 & 0\end{array}\right|-(0.344)\left|\begin{array}{cc}0.344 & 3.266 \\ 2.309 & 0\end{array}\right|+2.309\left|\begin{array}{cc}0.344 & 0.974-\mathrm{z} \\ 2.309 & 3.266\end{array}\right|$
$(1.218-z)[(0.974-z)(0)-(3.266)(3.266))-(0.344(-(3.266)(2.309))+$
$+(\mathbf{2 . 3 0 9})[(0.344)(3.266)-(0.974-z)(2.309))=0$
$\mathrm{Z}=-0.344$ (negative), Then $\mathrm{x}^{*}, \mathrm{y}^{*}$ is a relative maximum $f^{*}\left(\mathrm{x}^{*}, \mathrm{y}^{*}\right)=0.3248$

## Determinant Calculation

$$
A=\left[\begin{array}{lll}
1 & 2 & 3 \\
4 & 5 & 6 \\
8 & 8 & 9
\end{array}\right]
$$

$$
\begin{aligned}
\operatorname{det} A & =1\left|\begin{array}{ll}
5 & 6 \\
8 & 9
\end{array}\right|-2\left|\begin{array}{ll}
4 & 6 \\
8 & 9
\end{array}\right|+3\left|\begin{array}{ll}
4 & 5 \\
8 & 8
\end{array}\right| \\
& =-3+24-24
\end{aligned}
$$

$$
=-3
$$

$$
\left|\begin{array}{lll}
1 & 2 & 3 \\
4 & 5 & 6 \\
8 & 8 & 9
\end{array}\right|
$$

Solution. Using row operations $R_{2} \rightarrow R_{2}-4 R_{1}$ and $R_{3} \rightarrow R_{3}-8 R_{1}$ and then expanding along the first column, gives

$$
\begin{aligned}
& \left|\begin{array}{rrr}
1 & 2 & 3 \\
4 & 5 & 6 \\
8 & 8 & 9
\end{array}\right|=\left|\begin{array}{rrr}
1 & 2 & 3 \\
0 & -3 & -6 \\
0 & -8 & -15
\end{array}\right|=\left|\begin{array}{rr}
-3 & -6 \\
-8 & -15
\end{array}\right| \\
& =-3\left|\begin{array}{rr}
1 & 2 \\
-8 & -15
\end{array}\right|=\quad R_{2} \longrightarrow R_{2}+8 R_{1} \\
& =-3\left|\begin{array}{ll}
1 & 2 \\
0 & 1
\end{array}\right|=-3[1(1)-2(0)]=-3(1)=-3
\end{aligned}
$$

## http://matrix.reshish.com/determinant.php

## $\leftarrow \rightarrow$ C @ matrix.reshish.com/determinant.php



Show solution

Result:
Determinant is -3

## Necessary Conditions for a General Problem

The equations can be extended to the case of a general problem with $\boldsymbol{n}$ variables and $\boldsymbol{m}$ equality constraints:
Minimize $f(\mathbf{X})=f\left(\mathrm{x}_{1}, \mathrm{x}_{2}, \ldots \ldots, \mathrm{x}_{\mathrm{n}}\right)$
subject to: $\quad g_{j}(\mathbf{X})=0, \quad j=1,2, \ldots, m$
The Lagrange function, $L$, in this case is defined by introducing one Lagrange multiplier $\lambda_{j}$ for each constraint $g_{j}(X)$ as:

$$
L\left(x_{1}, x_{2}, \ldots, x_{n}, \lambda_{1}, \lambda_{2}, \ldots, \lambda_{m}\right)=f(\mathbf{X})+\lambda_{1} g_{1}(\mathbf{X})+\lambda_{2} g_{2}(\mathbf{X})+\cdots+\lambda_{m} g_{m}(\mathbf{X})
$$

The necessary conditions for the extremum of $L$, are given by:

$$
\frac{\partial L}{\partial x_{i}}=\frac{\partial f}{\partial x_{i}}+\sum_{j=1}^{m} \lambda_{j} \frac{\partial g_{j}}{\partial x_{i}}=0, \quad i=1,2, \ldots, n
$$

$$
\frac{\partial L}{\partial \lambda_{j}}=g_{j}(\mathbf{X})=0, \quad j=1,2, \ldots, m
$$

## Sufficient Condition

A sufficient condition for $f(X)$ to have relative minimum at $X^{*}$ is that the quadratic, $Q$, defined by:

$$
Q=\sum_{i=1}^{n} \sum_{j=1}^{n} \frac{\partial^{2} L}{\partial x_{i} \partial x_{j}} d x_{i} d x_{j}
$$

evaluated at $\mathrm{X}=\mathrm{X} *$ must be positive definite for all values of variations $d X$ for which the constraints are satisfied. If Q is negative definite for all choices of the admissible variations $d x_{i}, \mathrm{X}^{*}$ will be a constrained maximum of $f(\mathrm{X})$. It has been shown by Hancock that a necessary condition for the quadratic form $Q$, to be positive (negative) definite for all admissible variations dX is that each root of the polynomial Zi , defined by the following determinantal equation, be positive (negative):
$\left|\begin{array}{ccccccccc}L_{11}-z & L_{12} & L_{13} & \cdots & L_{1 n} & g_{11} & g_{21} & \cdots & g_{m 1} \\ L_{21} & L_{22}-z & L_{23} & \cdots & L_{2 n} & g_{12} & g_{22} & \cdots & g_{m 2} \\ \vdots & & & & & & & & \\ L_{n 1} & L_{n 2} & L_{n 3} & \cdots & L_{n n}-z & g_{1 n} & g_{2 n} & \cdots & g_{m n} \\ g_{11} & g_{12} & g_{13} & \cdots & g_{1 n} & 0 & 0 & \cdots & 0 \\ g_{21} & g_{22} & g_{23} & \cdots & g_{2 n} & 0 & 0 & \cdots & 0 \\ \vdots & & & & & & & & \\ g_{m 1} & g_{m 2} & g_{m 3} & \cdots & g_{m n} & 0 & 0 & \cdots & 0\end{array}\right|=0$

## Where:

$$
L_{i j}=\frac{\partial^{2} L}{\partial x_{i} \partial x_{j}}\left(\mathbf{X}^{*}, \lambda^{*}\right), g_{i j}=\frac{\partial g_{i}}{\partial x_{j}}\left(\mathbf{X}^{*}\right)
$$

This equation on expansion, leads to an ( $n-\mathrm{m}$ )th-order polynomial in $z$. If some of the roots of this polynomial are positive while the others are negative, the point $\mathrm{X}^{*}$ is not an extreme point.

### 1.7.2 Multivariable Optimization With Inequality Constraints

## Consider the following problem: Minimize $f(\mathbf{X})$

 subject to: $g_{j}(X) \leq 0, j=1,2, \ldots, m$
## Kuhn-Tucker Conditions

The conditions to be satisfied at a constrained minimum point, $\mathrm{X}^{*}$. These conditions are, in general, not sufficient to ensure a relative minimum. However, there is a class of problems, called convex programming problems for which the Kuhn-Tucker conditions are necessary and sufficient for a global minimum.

## Kuhn-Tucker Conditions

The Kuhn-Tucker conditions can be stated as follows:

$$
\begin{aligned}
\frac{\partial f}{\partial x_{i}}+\sum_{j=1}^{m} \lambda_{j} \frac{\partial g_{j}}{\partial x_{i}} & =0, \quad i=1,2, \ldots, n \\
\lambda_{j} g_{j} & =0, \\
& \\
g_{j} & \leq 0, \\
& j=1,2, \ldots, m \\
\lambda_{j} & \geq 0, \\
& j=1,2, \ldots, m \\
& j 1,2, \ldots, m
\end{aligned}
$$

Note that if the problem is one of maximization or if the constraints are of the type $g_{j} \geq 0$, the $\lambda_{\mathrm{j}}$ have to be nonpositive. On the other hand, if the problem is one of maximization with constraints in the form $g_{j} \geq 0$, the $\lambda_{\mathrm{j}}$ have to be nonnegative.

## Types of Nonlinear Programming

- Nonlinear objective function, linear constraints.
- Nonlinear objective function and nonlinear constraints.
- Linear objective function and nonlinear constraints.


## Nonlinear Objective Function and Linear Constraints:

The Great Western Appliance Company sells two models of toaster ovens, the Micro toaster $\left(X_{1}\right)$ and the Self-Clean Toaster Oven $\left(X_{2}\right)$.
The firm earns a profit of $\mathbf{\$ 2 8}$ for each Micro toaster regardless of the number sold. Profits for the Self-Clean model, however, increase as more units are sold because of fixed overhead. Profit on this model may be expressed as $21 X_{2}+0.25 X_{2}{ }^{2}$.
Great Western's profit is subject to two linear constraints on production capacity and sales unit time available.
Max: $28 X_{1}+21 X_{2}+0.25 X_{2}^{2}$
Subject to:
$X_{1}+\quad X_{2} \leq 1000$ (units of production capacity),
$0.5 X_{1}+0.4 X_{2} \leq 500$ (hours of sales time available).

$$
X_{1}, X_{2} \geq 0
$$

When an objective function contains a squared term and the problem constraints are linear, it is called a quadratic programming problem.

## An Excel Formulation of Great Western Appliance's Nonlinear Programming Problem.

## PROGRAM 10.9

Excel 2010 Solver Solution for Great Western Appliance NLP Problem

| - | A | B | c | D | E | F | G |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | Great Western Appliance |  |  |  |  |  |  |
| 2 |  | Micro | Self-Clean |  |  |  |  |
| 3 | Variables | X1 | X2 |  |  |  |  |
| 4 | Values | 0 | 1000 |  |  |  |  |
| 5 |  |  |  |  |  |  |  |
| 6 | Terms | $\mathrm{x}_{1}$ | X2 | X2 ${ }^{2}$ |  |  |  |
| 7 | Calculated Values | 0 | 1000 | 1000000 | Profit |  |  |
| 8 | Profit | 28 | 21 | 0.25 | 21000 |  |  |
| 9 |  |  |  |  |  |  |  |
| 10 | Constraints |  |  |  | LHS | Sign | RHS |
| 11 | Capacity | 1 | 1 |  | 1000 | $\leq$ | 1000 |
| 12 | Hours Available | 0.5 | 0.4 |  | 400 | $\leq$ | 500 |

## Key Formulas

| $A$ | $E$ |
| :---: | :---: |
| 8 | $=$ SUMPRODUCT(\$B\$7:\$D\$7,B8:D8) |
| 9 | LHS |
| 10 | $=$ SUMPRODUCT(\$B\$4:\$C\$4,B11:C11) |
| 11 | $=$ SUMPRODUCT(\$B\$4:\$C\$4,B12:C12) |

## Solver Parameter Inputs and Selections

Set Objective: E8
By Changing cells: B4:C4
To: Max
Subject to the Constraints:
E11:E12 <= G11:G12

Solving Method: GRG Nonlinear

- Make Variables Non-Negative


## Input Screen

$$
\begin{gathered}
\text { Max: } 28 X_{1}+21 X_{2}+0.25 X_{2}{ }^{2} \\
X_{1}+\quad X_{2} \leq 1000 \\
0.5 X_{1}+\mathbf{0 . 4} X_{2} \leq 500 \\
\hline
\end{gathered}
$$





## Both Nonlinear Objective Function and Nonlinear Constraints.

The annual profit at a medium-sized (200-400 beds) Hospital Corporation-owned hospital depends on the number of medical patients admitted $\left(\mathrm{X}_{1}\right)$ and the number of surgical patients admitted $\left(\mathrm{X}_{2}\right)$. The nonlinear objective function for Hospicare is:
Max. $13 X_{1}+6 X_{1} X_{2}+5 X_{2}+1 / X_{2}$.
The corporation identifies three constraints, two of which are also nonlinear, that affect operations. They are
$2 X_{1}^{2}+4 X_{2}^{2} \leq 90$ (nursing capacity, in thousands of labor-days).
$\mathrm{X}_{1}+\leq 75$ (x-ray capacity, in thousands).
$8 \mathrm{X}_{1}-X 2_{2} \mathrm{X}_{2} \leq 61$ (marketing budget required, in thousands of $\$)$.

Max: $\quad 13 X_{1}+6 X_{1} X_{2}+5 X_{2}+1 / X_{2}$
Subject to:

$$
\begin{array}{r}
2 X_{1}^{2}+4 X_{2}^{2} \leq 90 \\
X_{1}+X_{2}^{3} \leq 75 \\
8 X_{1}-2 X_{2} \leq 61
\end{array}
$$

## An Excel Formulation of Hospicare Corp.'s NLP Problem:




|  |  |  |  |  | ( |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 4 | A | B | C | D | E | F | G | H | I | J |
| 2 |  |  |  |  |  |  |  |  |  |  |
| 3 | Variables | X1 | X2 |  |  |  |  |  |  |  |
| 4 | Values | 6.066259 | 4.1 |  |  |  |  |  |  |  |
| 5 |  |  |  |  |  |  |  |  |  |  |
| 6 | Terms | X1 | X1 ${ }^{2}$ | X1 ${ }^{*}$ X2 | X2 | X2 ${ }^{3}$ | 1/X2 |  |  |  |
| 7 | Calculated Values | 6.066259 | 36.79949 | 24.87319 | 4.100253 | 68.93374 | 0.244 | Total Profit |  |  |
| 8 | Profit | 13 | 0 | 6 | 5 |  | 1 | 248.845671 |  |  |
| 9 |  |  |  |  |  |  |  |  |  |  |
| 10 | Constraints |  |  |  |  |  |  | LHS | Sign | RHS |
| 11 | Nursing |  | 2 |  | 4 |  |  | 89.999998 | $\leq$ | 90 |
| 12 | X-Ray | 1 |  |  |  | 1 |  | 75 | $\leq$ | 75 |
| 13 | Budget | 8 |  |  | -2 |  |  | 40.329564 | $\leq$ | 61 |
| 14 |  |  |  |  |  |  |  |  |  |  |

## Linear Objective Function with Nonlinear Constraints

Thermlock Corp. produces massive rubber washers and gaskets like the type used to seal joints on the NASA Space Shuttles. To do so, it combines two ingredients; rubber ( $\mathrm{X}_{1}$ ) and oil ( $\mathrm{X}_{2}$ ).
The cost of the industrial quality rubber used is $\$ 5$ per pound and the cost of the high viscosity oil is $\$ 7$ per pound. Two of the three constraints Thermlock faces are nonlinear. The firm's objective function and constraints are

Min. $5 X_{1}+7 X_{2}$
Subject to:
$3 X_{1}+0.25 X_{1}^{2}+4 X_{2}+0.3 X_{2}^{2} \geq 125$ (hardness constraint),
$13 X_{1}+X_{1}^{3} \geq 80$ (tensile strength),
$0.7 X_{1}+X_{2} \geq 17$ (elasticity).

## Excel Formulation of Thermlock's NLP Problem:



## Solution to Thermlock's NLP Problem Using Excel Solver:

| PMicrosoit Excel-capturesxis - |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| [3] Eile Edit View Insert Format Took Data QM Window Hep |  |  |  |  |  |  |  |  | - |
|  | A | B | C | D | E | F | G | H | \| |
| 1 Thermlock Gaskets | Thermlock Gaskets |  |  |  |  |  |  |  |  |
| 2 |  |  |  |  |  |  |  |  |  |
| 3 |  | $\times 1$ | X2 |  |  |  |  |  |  |
| 4 | value | 3.325326 | 14.67227 |  |  |  |  |  |  |
| 5 |  |  |  |  | total |  |  |  |  |
| 6 | cost | 5 | 7 |  | 119.3325 |  |  |  |  |
| 7 |  |  |  |  |  |  |  |  |  |
| 8 | constraints |  |  |  |  |  |  |  |  |
| 9 |  | $\times 1$ | $\times 1 \wedge 2$ | $\times 1 \times 3$ | x2 | X2^2 |  |  |  |
| 10 | value | 3.325326 | 11.05779 | 36.77076 | 14.67227 | 215.2756 | Total |  |  |
| 11 | Constraint 1 | 3 | 0.25 |  | 4 | 0.3 | $136.0122>$ |  | 125 |
| 12 | Constraint 2 | 13 |  | 1 |  |  | $80>$ |  | 80 |
| 13 | Constraint 3 | 0.7 |  |  | 1 |  | 17 > |  | 17 |

## Computational Procedures -Nonlinear Programming

Unlike LP methods:
One disadvantage of NLP is that the solution procedures to solve nonlinear problems do not always yield an optimal solution in a finite number of steps. The solution yielded may only be a local optimum, rather than a global optimum. In other words, it may be an optimum over a particular range, but not overall.
There is no general method for solving all nonlinear problems.

- Classical optimization techniques based on calculus, can handle some special cases, usually simpler types of problems.

Gradient method (steepest ascent method)

It is an interactive procedure that moves from one feasible solution to the next in improving the value of the objective function.
It has been computerized and can handle problems with both nonlinear constraints and objective.

## Separable programming

- Linear representation of nonlinear problem.
- Separable programming deals with a class of problems in which the objective and constraints are approximated by linear functions. In this way, the powerful simplex algorithm may again be applied.

In general, work in the area of NLP is the least charted and the most difficult of all the quantitative analysis models.

