

Numerical Optimization :- Unconstrained optimization problem :

Ref:- Numerical Optimization with Applications by
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In the previous lectures we have seen the techniques to solve ~~un~~ constrained optimization problems (like equality ~~&~~ constrained or inequality or mixed).

Now, we study some standard algorithms for unconstrained minimization of a functions of one variables as well as functions of several variables.

The problem formulation :-

$$\min_{x \in \mathbb{R}^n} f(x) \quad \text{---} \quad (*)$$

Defⁿ: Solution set :- Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be continuously differentiable function. Then the set

$\Omega = \{ x \in \mathbb{R}^n ; \nabla f(x) = 0 \}$ is called the solution set of $(*)$.

Remark :- Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be differentiable convex function and \bar{x} be a point of the solution set. Then \bar{x} is called the global minima of problem $(*)$.

Before we present any specific algorithm for solving the problem $(*)$, we list certain desirable properties which we ideally expect the given algorithm to process.

Defⁿ: (Descent Property)

An algorithm for solving the unconstrained minimization problem $(*)$ is said to have the descent property if the objective function value decreases as we go through the sequence $\{x^{(k)}\}$, i.e.

$$f(x^{(k+1)}) < f(x^{(k)}) \quad \forall k.$$

Globally Convergent :- An algorithm for the unconstrained minimization problem $(*)$ is said to be globally convergent if ~~for~~ starting from any point $x^{(0)} \in \mathbb{R}^n$, the sequence $\{x^{(k)}\}$ always converges to a point of the solution set Ω .

Defⁿ: (Order of Convergence) :-

(2)

Let the sequence $\{x^{(k)}\}$ converge to a point x^* and let $x^{(k)} \neq x^*$ for sufficiently large k . The quantity $\|x^{(k)} - x^*\|$ is called the error of the k^{th} iterate $x^{(k)}$. Suppose that there exists p and $0 < \alpha < \infty$ such that

$$\lim_{k \rightarrow \infty} \frac{\|x^{(k+1)} - x^*\|}{\|x^{(k)} - x^*\|^p} = \alpha, \quad (0 < \alpha < \infty)$$

then p is called the order of convergence of the sequence $\{x^{(k)}\}$. Thus

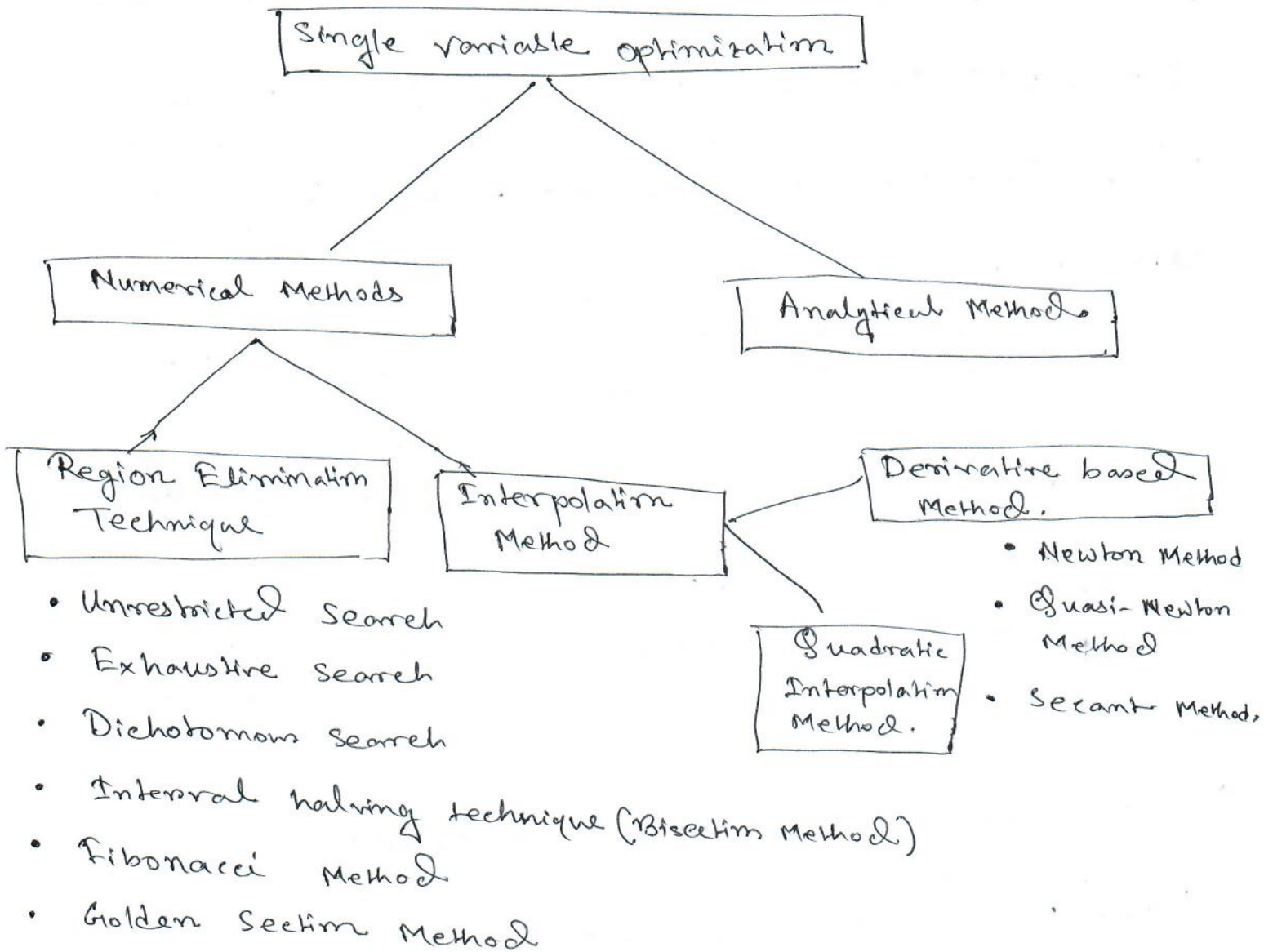
$$\|x^{(k+1)} - x^*\| = \alpha \|x^{(k)} - x^*\|^p \text{ asymptotically,}$$

Remark :- (i) If $p = 1$ the seqⁿ $\{x^{(k)}\}$ is said to have linear convergence rate.

(ii) If $p = 2$, it is called quadratic convergence rate.

Numerical Optimization Technique.

(NLP, with unconstrained)

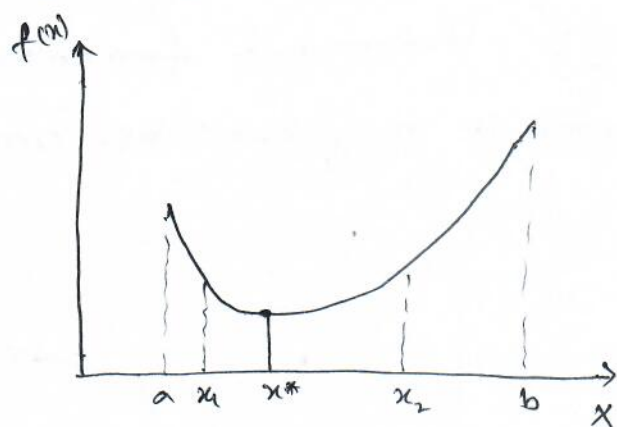
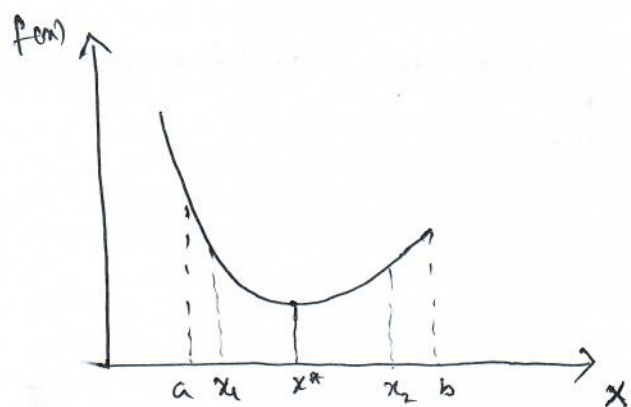


■ To use Region Elimination Technique the function should ⁽³⁾ be unimodal in the specified domain of definition.

Defⁿ :- Unimodal function :-

A function $f(x)$ is said to be unimodal in $a \leq x \leq b$ if it is monoton on either side, x^* is optimal point within the interval

- (i) if $f(x_1) > f(x_2) \Rightarrow x^* \in [x_1, b)$ but $x^* \notin (a, x_1)$
- (ii) if $f(x_1) < f(x_2) \Rightarrow x^* \in [a, x_2)$ but $x^* \notin (x_2, b]$.
- (iii) if $f(x_1) = f(x_2) \Rightarrow x^* \in [x_1, x_2]$



Remark:- f is called strictly unimodal if the function is unimodal and there does not exist any sub interval within the range $[a, b]$, where the function is having constant value.

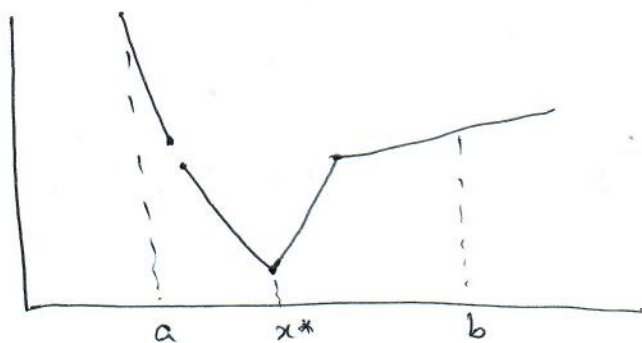
Remark:- Note that we do not need explicit knowledge of the function f (continuity or differentiability) as long as it is possible to get its value at a specified point.

Alternate definition of Unimodal function:

The function $f : [a, b] \rightarrow \mathbb{R}$ is said to be a unimodal (to be specific unimodal min) function if it has only one mode i.e. it has a single relative min, i.e. $\exists a \leq x^* \leq b$ such that

- (i) f is strictly decreasing (\downarrow) in $[a, x^*)$
- (ii) f is strictly increasing (\uparrow) in $[x^*, b]$

Note :- Unimodal function may not be differentiable, even it may not be continuous.



Region Elimination Technique.

(1)

$$\text{Min } f(x), \quad a \leq x \leq b.$$

Assumption: f is unimodal function and x^* is min point $a \leq x^* \leq b$.

Step-1: Starting with the initial guess point,
 $a < x_1 < x_2 < b$.

(i) if $f(x_1) > f(x_2)$ then minimum must lie within (x_1, b)

(ii) if $f(x_1) < f(x_2)$ then minimum must lie within (a, x_2) .

(iii) if $f(x_1) \approx f(x_2)$ then minimum must lie (x_1, x_2) ,
i.e. eliminate the intervals (a, x_1) and (x_2, b) .

