# Optimization 

## 1-D optimization

Chapter 13:

## Optimization

- Often in engineering we first learn how to design units or systems that simply meet a set of specifications
- In practical application, the design must also be optimized to:
- Reduce: cost, raw material, waste, downtime, environmental impact, inventory
- Increase: profits, yields, safety, reliability


## Optimization

- So the trick is to formulate a function to be optimized
- For a given operation, costs can be minimized
- For a given supply, product can be maximized
- For an existing process, recycle ratios, purge rates, temperatures, etc can be manipulated to increase yeild


## Optimization

- Objective function
- goal of the optimization
- Design variables
- Which independent variables can be manipulated to achieve the goal
- Constraints
- External limitations of the system
- Often in terms of inequalities
- Limitation of the range of the desired solution (e.g. no negative temperatures)


## One Dimensional Optimization

- Minimizing or maximizing a function of a single independent variable is quite common in engineering design, especially in unit ops
- Common methods include:
- Newton's method, if the function is differentiable
- Quadratic interpolation, which is analogous to Müller's Method and the Secant method
- Golden-section search, which is analogous to the bracketed bisection method


## Golden Section Search

This method starts at to initialization points, $x_{l}$ and $x_{u}$
Then these and two interior points, $x_{1}$ and $x_{2}$, are all evaluated

Since we are looking for the maximum, the lowest of the values $f\left(x_{1}\right), f\left(x_{2}\right)$ determines which side of the interval to continue with

Assuming you have the optimum bracketed, this method will always find it
It is similar to the bisection method in that it converges predictably although slightly slower



Second
iteration
$\leftarrow \ell_{2} \rightarrow$

## Golden Section Search

The two interior points are chosen carefully, so that the ratio of the intervals are equal
This allows us to 'recycle' the function values and reduce computations
Only one new x value per iteration



## Golden Section Search

Example 13.1
Use the golden-section search to find the maximum of the function $\mathrm{f}(\mathrm{x})$ within the interval $X_{L}=0$ and $X u=4$.

$$
f(x)=2 \sin x-\frac{x}{10}^{2}
$$

- Start at $x=0,4$ and calculate $d=G R^{*}(4-0)$
- Calculate $\mathrm{x} 1, \mathrm{x} 2$ and $\mathrm{f}(\mathrm{x} 1), \mathrm{f}(\mathrm{x} 2)$
- Throw away xl or xu
- Replace x1 or x2


## Golden Section Search

|  | A | B | C | D | E | F | G | H | 1 | J | K | L |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | Example 13.1 |  |  | Ratio | 0.618034 |  | Solver Solution |  |  |  |  |  |
| 2 |  |  |  |  |  |  | x_max | 1.427551 | $f(x)$ | 1.775726 |  |  |
| 3 | i | x | $\mathrm{f}(\mathrm{xl})$ | $\times 1$ | $f(x 1)$ | x2 | $\mathrm{f}(\mathrm{x} 2)$ | Xu | $f(x u)$ | d | Throw away | max |
| 4 | 1 | 0.00000 | 0.0 (1) 00 | 1.52786 | 1.76472 | 2.47214 | 0.62997 | 4.00000 | -3.11360 | 2.4721 | upper | 1.76472 |
| 5 | 2 | 0.00000 | 0.00000 | 0.94427 | 1.53098 | 1.52786 | 1.76472 | 2.47214 | 0.62997 | 1.5279 | lower | 1.76472 |
| 6 | 3 | 0.94427 | 1.53098 | 1.52786 | 1.76472 | 1.88854 | 1.54322 | 2.47214 | 0.62997 | 0.9443 | upper | 1.76472 |
| 7 | 4 | 0.94427 | 1.53098 | 1.30495 | 1.75945 | 1.52786 | 1.76472 | 1.88854 | 1.54322 | 0.5836 | lower | 1.76472 |
| 8 | 5 | 1.30495 | $1.75945^{\prime}$ | 1.52786 | 1.76472 | 1.66563 | 1.71358 | 1.88854 | 1.54322 | 0.3607 | upper | 1.76472 |
| 9 | 6 | 1.30495 | $1.75945^{\prime}$ | 1.44272 | 1.77547 | 1.52786 | 1.76472 | 1.66563 | 1.71358 | 0.2229 | upper | 1.77547 |
| 10 | 7 | 1.30495 | $1.75945^{\prime}$ | 1.39010 | 1.77420 | 1.44272 | 1.77547 | 1.52786 | 1.76472 | 0.1378 | lower | 1.77547 |
| 11 | 8 | 1.39010 | 1.77420 | 1.44272 | 1.77547 | 1.47524 | 1.77324 | 1.52786 | 1.76472 | 0.0851 | upper | 1.77547 |
| 12 | 9 | 1.39010 | 1.77420 | 1.42262 | 1.77570 | 1.44272 | 1.77547 | 1.47524 | 1.77324 | 0.0526 | upper | 1.77570 |
| 13 | 10 | 1.39010 | 1.77420 | 1.41020 | 1.77540 | 1.42262 | 1.77570 | 1.44272 | 1.77547 | 0.0325 | lower | 1.77570 |
| 14 | 11 | 1.41020 | 1.77540 | 1.42262 | 1.77570 | 1.43030 | 1.77572 | 1.44272 | 1.77547 | 0.0201 | lower | 1.77572 |
| 15 | 12 | 1.42262 | 1.77570 | 1.43030 | 1.77572 | 1.43504 | 1.77566 | 1.44272 | 1.77547 | 0.0124 | upper | 1.77572 |
| 16 | 13 | 1.42262 | 1.77570 | 1.42736 | 1.77573 | 1.43030 | 1.77572 | 1.43504 | 1.77566 | 0.0077 | upper | 1.77573 |
| 17 | 14 | 1.42262 | 1.77570 | 1.42555 | 1.77572 | 1.42736 | 1.77573 | 1.43030 | 1.77572 | 0.0047 | lower | 1.77573 |
| 18 | 15 | 1.42555 | 1.77572 | 1.42736 | 1.77573 | 1.42848 | 1.77572 | 1.43030 | 1.77572 | 0.0029 | upper | 1.77573 |
| 19 | 16 | 1.42555 | 1.77572 | 1.42667 | 1.77572 | 1.42736 | 1.77573 | 1.42848 | 1.77572 | 0.0018 | lower | 1.77573 |
| 20 | 17 | 1.42667 | 1.77572 | 1.42736 | 1.77573 | 1.42779 | 1.77573 | 1.42848 | 1.77572 | 0.0011 | upper | 1.77573 |
| 21 | 18 | 1.42667 | 1.77572 | 1.42710 | 1.77573 | 1.42736 | 1.77573 | 1.42779 | 1.77573 | 0.0007 | lower | 1.77573 |
| 22 | 19 | 1.42710 | 1.77573 | 1.42736 | 1.77573 | 1.42753 | 1.77573 | 1.42779 | 1.77573 | 0.0004 | lower | 1.77573 |
| 23 | 20 | 1.42736 | 1.77573 | 1.42753 | 1.77573 | 1.42763 | 1.77573 | 1.42779 | 1.77573 | 0.0003 | upper | 1.77573 |
| 24 | 21 | 1.42736 | 1.77573 | 1.42747 | 1.77573 | 1.42753 | 1.77573 | 1.42763 | 1.77573 | 0.0002 | lower | 1.77573 |
| 25 | 22 | 1.42747 | 1.77573 | 1.42753 | 1.77573 | 1.42757 | 1.77573 | 1.42763 | 1.77573 | 0.0001 | lower | 1.77573 |
| 26 | 23 | 1.42753 | 1.77573 | 1.42757 | 1.77573 | 1.42759 | 1.77573 | 1.42763 | 1.77573 | 0.0001 | upper | 1.77573 |
| 27 | 24 | 1.42753 | 1.77573 | 1.42755 | 1.77573 | 1.42757 | 1.77573 | 1.42759 | 1.77573 | 0.0000 | upper | 1.77573 |
| 28 |  |  |  |  |  |  |  |  |  |  |  |  |

## Twenty interations to get error less than four decimal places

## Quadratic Interpolation

For three points there is one parabola that can be fitted to the points (like Müller's method)

- We can analytically determine the location of the maximum of this parabola
Then evaluate the function at this location and determine a new range to consider



## Quadratic Interpolation

- The selection criteria for which endpoint to discard, is similar to Golden Search
- But still only one new function evaluation is required per iteration
- Although this can be a bracketing method, it will work as an open method as well
- Questions?



## Newton's Method

- Basically finding the root of the derivative
- When the derivative equals zero a minimum or maximum has been found
- Find the roots of $g(x)=f$ ' $(x)$ using methods discussed in part 3
- Then plug $x$ into $f(x)$ to find the value at the maximum
- Questions?


## Summary

- The methods discussed today have the same weaknesses and strengths of their analogous methods for roots
- Golden Section
- Must know the range containing a single optimum point
- Slow
- Quadratic Interpolation
- Like false-position, this method can get hung up on one side of the range and slowly converge
- Newton's
- Need first and second derivative
- Must be 'close' with intial guesses
- Fast convergence

