

Operations Research Techniques and Algorithms (620-361)

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Today's Lecture

Introduction

Unimodal 1-D unconstrained optimisation

Fibonacci search

Golden section, false position and Newton's method

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In step 1 of this algorithm, we need to find the smallest value of n such that $(b - a)/F_n < 2\epsilon$. We could do this by generating numbers F_n using the recursive formula, until $F_n > (b - a)/2\epsilon$.

An alternative to generating the numbers F_n using the recursive formula, is to use the explicit formula for F_n .

This is derived as follows.

Note that

$$F_n = F_{n-1} + F_{n-2}. \quad (1)$$

is a difference equation with constant coefficients. We solve these by trying a solution of the form $F_n = \lambda^n$. This leads to the *characteristic equation*

$$\lambda^2 = \lambda + 1 \quad (2)$$

and so

$$\lambda = \frac{1 \pm \sqrt{5}}{2}. \quad (3)$$

Thus

$$F_n = A \left(\frac{1 + \sqrt{5}}{2} \right)^n + B \left(\frac{1 - \sqrt{5}}{2} \right)^n$$

The initial conditions give us

$$A = \frac{\sqrt{5} + 1}{2\sqrt{5}}$$

and

$$B = \frac{\sqrt{5} - 1}{2\sqrt{5}},$$

so that

$$F_n = \frac{1}{\sqrt{5}} \left[\left(\frac{1 + \sqrt{5}}{2} \right)^{n+1} - \left(\frac{1 - \sqrt{5}}{2} \right)^{n+1} \right]. \quad (4)$$

The Golden Section Search

The Fibonacci Search reduces the size of an interval in which the minimum occurs in the most efficient way possible. However, at iteration k , we have to choose the points q and p so that the ratios

$$(q - a)/(b - a) = (b - p)/(b - a) = F_{k-1}/F_k$$

depend on k , and the entire search therefore depends on n . Also, we have to pre-compute the values of the Fibonacci numbers in some way.

To remove these dependencies, we can approximate the Fibonacci Search as follows: choose the position of q and p so that the above ratio is equal to

$$\gamma = \lim_{k \rightarrow \infty} F_{k-1}/F_k.$$

This is called the Golden Section search.

The Golden Section Search Let $\gamma_k = F_{k-1}/F_k$. Then, from (4),

$$\begin{aligned}\gamma_k &= \frac{\left(\frac{1+\sqrt{5}}{2}\right)^k - \left(\frac{1-\sqrt{5}}{2}\right)^k}{\left(\frac{1+\sqrt{5}}{2}\right)^{k+1} - \left(\frac{1-\sqrt{5}}{2}\right)^{k+1}} \\ &= \frac{1 - \left(\frac{1-\sqrt{5}}{1+\sqrt{5}}\right)^k}{\frac{1+\sqrt{5}}{2} - \frac{1-\sqrt{5}}{2} \left(\frac{1-\sqrt{5}}{1+\sqrt{5}}\right)^k}.\end{aligned}\tag{5}$$

Since $|1 - \sqrt{5}| < |1 + \sqrt{5}|$, the second terms in both the denominator and numerator decay as $k \rightarrow \infty$. Thus

$$\gamma \equiv \lim_{k \rightarrow \infty} \gamma_k \quad (6)$$

$$= \frac{1}{\frac{1+\sqrt{5}}{2}}$$

$$= \frac{\sqrt{5} - 1}{2}$$

$$\approx 0.618$$

(7)

We could also note that, by equation (1),

$$\gamma_k + \gamma_k \gamma_{k-1} = 1.$$

Letting $k \rightarrow \infty$, we see that

$$\gamma + \gamma^2 = 1. \tag{8}$$

The only point in the interval $[0, 1]$ that satisfies this equation is $(\sqrt{5} - 1)/2$, which agrees with our calculation above. The interesting thing about equation (8) is that it shows us that the ratio of γ to 1 is the same as the ratio of $1 - \gamma$ to γ . The number γ is known as the *golden ratio*, a ratio that is supposed to possess many aesthetic properties.

Golden Section Search Algorithm

To minimise a unimodal function f over $[a, b]$ to within tolerance ϵ .

1. Set

$$k = 1$$

$$p = b - \gamma(b - a)$$

$$q = a + \gamma(b - a)$$

Calculate $f(p)$ and $f(q)$.

2. Set $k = k + 1$. If $f(p) \leq f(q)$, then set

$$b = q$$

$$q = p$$

$$p = b - \gamma(b - a)$$

Calculate $f(p)$. If $f(p) > f(q)$, then set

$$a = p$$

$$p = q$$

$$q = a + \gamma(b - a)$$

Calculate $f(q)$. Repeat until $(b - a) < 2\epsilon$.

Methods Using the Derivative

Now let us assume not only that f is continuous, but that it is also differentiable and that we have a “black box” for f' as well as one for f . We still assume that f is unimodal on $[a, b]$.

The problem of minimising f now reduces to the problem of finding the point x^* , where $f'(x^*) = 0$.

The Method of False Position

Let g be a continuous and increasing function on \mathbb{R} . The method of false position is a recursive method for finding the point x^* where $g(x^*) = 0$.

Assume that we can find two points a and b such that $g(a) < 0$ and $g(b) > 0$. The line through the points $(a, g(a))$ and $(b, g(b))$ has equation

$$y = \frac{g(b) - g(a)}{b - a}(x - a) + g(a). \quad (9)$$

The value of the right hand side of (9) is equal to zero at the point

$$x_{estimate}^* = a + \frac{(b - a)g(a)}{g(a) - g(b)}. \quad (10)$$

$x_{estimate}^*$ is a linear estimate for the point x^* at which $g(x^*) = 0$.

- ▶ If $g(x_{estimate}^*) = 0$ (to within some tolerance), then we have found a “near enough” approximation to the point x^* .
- ▶ If $g(x_{estimate}^*) < 0$, then we use $x_{estimate}^*$ as the lower bound for a new interval.
- ▶ If $g(x_{estimate}^*) > 0$, then we use $x_{estimate}^*$ as the upper bound for a new interval.

We can extend this to develop a recursive algorithm as follows.

Algorithm for the Method of False Position

For an increasing, continuous function g on $[a, b]$, to find a point x^* where $|g(x^*)| < \epsilon$.

1. Set

$$k = 1$$
$$p = a + \frac{(b-a)g(a)}{g(a) - g(b)}$$

Calculate $g(p)$.

2. Set $k = k + 1$.

If $g(p) < 0$, then set

$$a = p$$

$$p = a + \frac{(b - a)g(a)}{g(a) - g(b)}$$

Calculate $g(p)$.

2. If $g(p) > 0$, then set

$$b = p$$
$$p = a + \frac{(b-a)g(a)}{g(a) - g(b)}$$

Calculate $g(p)$.

Repeat until $|g(p)| < \epsilon$.

If f is convex, then f' is increasing and so we can use the algorithm for the method of false position with $g = f'$ to find the minimum of f .

Newton's Method (Single Variable case)

Another method that we can use to find the point at which an increasing function g is equal to zero is Newton's Method.

The method is both simple to implement and typically converges very quickly. For this method, we need black boxes for both the function g and its derivative g' .

The underlying idea of Newton's Method is that we calculate an estimate of the point x^* at which $g(x^*) = 0$ by calculating the point at which the tangent to g at some point a would cross the x -axis.

The tangent to g at the point $(a, g(a))$ has slope $g'(a)$, and so its equation is

$$y = g'(a)(x - a) + g(a). \quad (11)$$

Thus $y = 0$ when $x = a - g(a)/g'(a)$.

We incorporate this into a recursive algorithm by calculating the value of $g(x)$ and $g'(x)$ at then a subsequent new estimate for the point at which $g = 0$.

One thing we need to beware of is that we do not “divide by zero”. Thus we need to test whether $g'(a)$ is very small. In fact if there is a zero of g' close to the zero of g , then Newton's Method doesn't work. Can you explain this physically?

Algorithm for Newton's Method

For an increasing, continuous function g on \mathbb{R} and an initial starting point a , to find a point x^* where $|g(x^*)| < \epsilon$.

1. Set

$$k = 1$$

if $g'(a) < \epsilon$ then signal and stop.

$$\text{else } p = a - \frac{g(a)}{g'(a)}$$

2. Set $k = k + 1$.

$$\begin{aligned} & a = p \\ \text{if } & g'(a) < \epsilon \text{ then signal and stop.} \\ \text{else } & p = a - \frac{g(a)}{g'(a)} \end{aligned}$$

Repeat until $|g(p)| < \epsilon$.

As with the method of false position, we can find the minimum of a convex and twice differentiable function f by using Newton's Method with $g = f'$ and $g' = f''$.