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BEST EXPLORATION FOR MAXIMUM IS FIBONACCIAN,

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SUMMARY

A unimodal function of one variable is defined on an interval. No regularity conditions involving continuity, derivatives, etc., are assumed. We wish to minimize the number of calculations of values of the function in order to assure the location of its maximum to a prescribed degree of accuracy. The solution of this problem and its discrete analogue involves the well-known Fibonacci sequence.
BEST EXPLORATION FOR MAXIMUM IS FIBONACCIAN

Selmer M. Johnson

When the author obtained the result presented here, he thought it was new. Subsequently, it was found that the result had previously been published by J. Kiefer [1]. However, the simplicity of the present proof may justify a short note on the problem.

In some practical problems the only way to locate a maximum of a function is to observe the values of the function at various points and compare them. For example, the function may not be continuous, or perhaps the derivatives may be too hard to find analytically, etc. With this in mind, and as a first step toward the solution of the general problem, we consider the case of a unimodal function on an interval.

Definition: A function $f(x)$ is unimodal if there is an $x_0$ such that $f$ is either strictly increasing for $x < x_0$ and strictly decreasing for $x > x_0$, or else strictly increasing for $x < x_0$ and strictly increasing for $x > x_0$.

For example, concave functions are unimodal.

In [1], Kiefer considers the problem of determining an interval containing the point at which a unimodal function on the unit interval possesses a maximum. No regularity conditions
concerning continuity or derivatives, etc., are assumed. He gives, for every \( \varepsilon > 0 \) and every number \( n \) of values of the argument at which the function may be observed, a procedure which is \( \varepsilon \)-minimax among the class of all sequential non-randomized procedures which terminate by giving an interval containing the required point, where the payoff is the length of the final interval.

We present a short proof of the result after it has been restated as follows.

**Theorem 1.** Let \( y = f(x) \) be any unimodal function defined on an interval \( 0 \leq x \leq L \). Let \( F_n = \sup \) of all \( L_n \) with the property that we can *always* locate the maximum of \( f(x) \) on a unit-length subinterval by calculating \( n \) values of the function. Then \( F_n \) is the \( n \)-th Fibonacci number; that is,

\[
F_n = F_{n-1} + F_{n-2}, \quad n \geq 2,
\]

with \( F_0 = F_1 = 1 \).

The proof will be by induction.

We observe first that \( F_0 = F_1 = 1 \) follows from the definition of \( F_n \), since just one value of the function gives no information concerning the location of the maximum.

Fix \( n \geq 2 \) and calculate \( y_1 = f(x_1), y_2 = f(x_2) \), where \( 0 < x_1 < x_2 < L \). If \( y_1 > y_2 \), the maximum occurs on \((0, x_2)\) since \( f(x) \) is strictly unimodal. If \( y_2 > y_1 \), the maximum is
on \((x_1, L_n)\). If \(y_1 = y_2\), choose either of the above intervals, even though we know the maximum occurs on \((x_1, x_2)\). Thus at each stage after the first computation we are left with a sub-interval and the value of \(f(x)\) at some interior point \(x\). Since values at the ends of an interval do not always give usable information relative to our problem, we restrict our attention to the interior points.

For \(n = 2\), \(L_n = 2 - \varepsilon\), take \(x_1 = 1 - \varepsilon\), \(x_2 = 1\), for arbitrarily small \(\varepsilon > 0\). By the argument of the preceding paragraph, we have \(F_2 = F_1 + F_0\).

Fix \(n > 2\), and assume that \(F_k = F_{k-1} + F_{k-2}\) for \(k = 2, \ldots, n - 1\). We shall show that

\[ F_n = F_{n-1} + F_{n-2}. \]

For if we calculate at \(x_1\) and \(x_2\) on \((0, 1)\) we have the picture

\[ \begin{array}{c}
0 \quad x_1 \quad x_2 \quad L_n \\
| | \\
y_1 \quad y_2
\end{array} \]

If \(y_1 > y_2\), we have the new picture

\[ \begin{array}{c}
0 \quad x_1 \quad x_2 \\
| \\
y_1
\end{array} \]

But then \(x_2 < F_{n-1}\) since we have only \(n - 2\) more choices with \(x_1\) a first choice for the case \(k = n - 1\). Moreover,
$x_1 < F_{n-2}$, since the maximum could occur on $(0, x_1)$ with two choices of $x$ already used.

Similarly if $y_2 > y_1$, we have $L_n - x_1 < F_{n-1}$. Thus

$L_n < F_{n-1} + x_1 < F_{n-1} + F_{n-2}$, so that $F_n < F_{n-1} + F_{n-2}$.

Now choose $L_n + (1 - \frac{\varepsilon}{2})(F_{n-1} + F_{n-2})$, $x_1 + (1 - \frac{\varepsilon}{2})F_{n-2}$, and $x_2 = (1 - \frac{\varepsilon}{2})F_{n-1}$. Since $\varepsilon$ is arbitrarily small this shows that $F_n = F_{n-1} + F_{n-2}$, so Theorem 1 is proved. Moreover, the procedure is optimal for a given $\varepsilon$, since after comparing $I(x_i)$ and $I(x_j)$ we are left with an interval of length $L_{n-1} = (1 - \frac{\varepsilon}{2})F_{n-1}$, and with a value at optimal first position for this smaller interval. Continuing in this way we have $L_k = (1 - \frac{\varepsilon}{2})F_k$ for $k \leq n$; in particular, $L_2 = (1 - \frac{\varepsilon}{2})F_2 = \frac{\varepsilon}{2}$, and the final interval is of unit length.

The $F_n$ form the sequence

$$1, 1, 2, 3, 5, 8, 13, 21, 34, 55, \ldots,$$

with $F_{20} > 10000$. Thus a maximum can always be located within $10^{-4}$ of the original interval length with 20 calculations. Using the more conventional technique of computing $f(x)$ and $f(x + \varepsilon)$, where $x$ is at the midpoint of the interval, may take as many as 28 computations.

As Kiefer remarks, since $F_{n+1}/F_n$ rapidly approaches its limit $\frac{1}{2}(\sqrt{5} + 1) = 1.618$, one could choose the first two values of $x$ at $L/\tau = 0.618 L$ from either end of the interval. This would be an excellent approximation at each stage except for the
last few choices, especially if \( n \) is not fixed in advance but is determined after several values have been observed (e.g., more observations might be taken if the function appears to be sharply peaked near its maximum). This constant ratio to determine the next \( x \) would be useful in a computing-machine code.

A line segment is said to be divided according to the "golden section" if one part is \( \tau \) times the other. The approximation suggested above gives at each stage an interval which is divided by the golden section at its \( x \) value but reduced in scale from the previous interval. This follows from the relation

\[
\frac{L / \tau}{L(1 - \frac{1}{\tau})} = \frac{1}{\tau - 1} = \tau .
\]

The Discrete Case

**Theorem 2.** Let \( y = f(x) \) be any unimodal function defined on a discrete set of \( H_n \) points. Let \( k_n = \text{max } H_n \) such that the function's maximum can always be identified in \( n \) observations. Then

\[ k_n = -1 + F_{n+1}, \quad n \geq 1 .\]

**Proof.** Number the points 1, 2, 3, \( \ldots \), \( H_n \). Observe that \( K_1 = 1, K_2 = 2, K_3 = 4 \). Fix \( n \geq 3 \) and assume \( K_k = -1 + F_{k+1} \) for \( k < n \). Calculate \( f(x) \) at \( x_1 \) and \( x_2 \). By arguments analogous
to those of Theorem 1, we must have

\[ x_1 \leq K_{n-2} + 1, \quad H_n - x_1 \leq K_{n-1}, \]

so that

\[ H_n \leq K_{n-2} + 1 + K_{n-1} = (F_{n-1} - 1) + 1 + F_n - 1 = F_{n+1} - 1. \]

This maximum is attained when \( x_1 = F_{n-1} \) and \( x_2 = F_n \). Thus Theorem 2 is proved; moreover, the optimal construction is given.
REFERENCE
