The Binary Search revisited

The Binary Search is a beautiful, simple, efficient, and—hence—well-known little algorithm for searching. Most computing scientists and programmers will be familiar with it. Unfortunately, there are two observations that cast some doubt on the level of familiarity. First, many programmers still have a relatively hard time writing down correct program code for it, in spite of the fact that it is an algorithm of just a few lines. Second, most programmers believe that the algorithm only works for searching in sorted arrays, which reveals a misunderstanding. (A long time ago, we ourselves, too, used to tell the binary search to our students, by first drawing an analogy with searching for a word in a dictionary. This we have recognized as a major educational blunder.) The purpose of this note is to remedy the situation.

* * *

Our starting point is that we are given a function or an array \( f[a..b] \), \( a < b \), of elements of some kind, such that the two outer elements \( f[a] \) and \( f[b] \) are in some relation \( Z \) to each other, a fact that we denote by \( a Z b \). The problem is to find a pair of neighbouring elements of \( f \) that are in the relation \( Z \). (Of course, such a pair need not exist; we will address this problem later on.) More precisely, we wish to construct a program with postcondition

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\[ R: \quad a \leq x \land x < b \land x \leq Z(x+1) \]
on the premise that the precondition implies
\[ a < b \land a \leq Z b. \]
The procedure for constructing such a program is quite standard. In view of the shapes of the pre- and the postcondition we try to establish \( R \) by means of a repetition with invariants \( P_0 \) and \( P_1 \), given by
\[ P_0: \quad a \leq x \land x < y \land y \leq b \]
\[ P_1: \quad x \leq Z y. \]
and with bound function \( y-x \). Our program thus gets the form
\[
\{ a < b \land a \leq Z b \}
\begin{array}{l}
x, y := a, b \\
\{ \text{inv } P_0 \land P_1 \} \{ \text{bnd } y-x \} \\
\text{do } y \neq x+1 \rightarrow \text{shrink } y-x \text{ under} \\
\text{invariance of } P_0 \land P_1 \\
\text{od} \\
\{ R \}
\end{array}
\]
And indeed, by construction, \( P_0 \land P_1 \land y = x+1 \Rightarrow R \).

For the shrinking of \( y-x \) in the body of the repetition we investigate an increase of \( x \) and a decrease of \( y \). (We do so since our problem is highly symmetric in \( x \) and \( y \).) On the assumption that \( x < b \), for some \( b \) still to be determined, assignment \( x := b \) increases \( x \). We now find out under what circumstances this
assignment maintains $P_0$ and $P_1$. As for the invariance of $P_0$, we observe

$$P_0 (x:=h)$$

$$\equiv \begin{cases} \text{definition of } P_0 \\ a \leq h \land h < y \land y \leq b \end{cases}$$

$$\equiv \begin{cases} P_0 \text{ and } x<h \text{ are preconditions to } x:=h, \\
\text{hence } a < h \land y \leq b \end{cases}$$

$$h < y.$$  

So we require $h$ to satisfy not just $x<h$ but also $h<y$, and hence

$$x < h \land h < y.$$  

**Note** Fortunately, such an $h$ exists because from $P_0$ and the guard $y \neq x+1$ of the repetition we conclude that $x+1 \leq y$.  
**End of Note.**

Having settled the invariance of $P_0$ under $x:=h$, we now turn our attention to $P_1$. We observe

$$P_1 (x:=h)$$

$$\equiv \begin{cases} \text{definition of } P_1 \\ h \leq y \end{cases}$$

Since this latter condition does not in general follow from $P_1$ or $P_0$, we plug it in as a guard to $x:=h$.

We next address a decrease of $y$. Because we already have an $h$ such that $h<y$, we propose $y:=h$. By the problem's symmetry in $x$ and $y$, our program now becomes
\{ a < b \land a Z b \}
\forall x, y \colon a, b
\{ inv \; P_0 \land P_1 \} \{ bnd \; y - x \}
; \; do \; y \neq x + 1 \rightarrow
\{ x + 2 \leq y \}
"h such that \; x < h \land h < y"
; \; if \; h Z y \rightarrow x = h
\forall x \; Z h \rightarrow y = h
\fi
\od
\{ \mathcal{R} \}

Prog 0

Remains the question of termination of this program. It does terminate if we can ensure that at least one of the guards of the if-statement evaluates to true. This will in general not be the case, but it is if relation \( Z \) satisfies

\( (0) \quad x Z y \Rightarrow h Z y \land x Z h \) \quad (\forall x, y, h)

(The antecedent is the valid precondition \( P_1 \); the consequent is the disjunction of the guards.)

While adopting this condition on \( Z \), we can even ensure very fast termination by choosing \( h \) equal to \((x + y) \div 2\), which thanks to precondition \( x + 2 \leq y \) indeed establishes \( x < h \land h < y \). And thus, binary search is born!

* * *

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Aside  Meanwhile we have proven an — admittedly not too deep — mathematical
theorem, viz that
\[ a < b \land a \leq b \] (the program's precondition)
and
\[ Z \text{ satisfies } (0) \]
there exists an \( x \) such that
\[ a \leq x \land x < b \land x \leq Z(x+1) \] (the program's postcondition).

It is a discrete analogue of the well-known mean value theorem for continuous functions on a closed interval.

End of Aside.

Our program computes an \( x \) satisfying postcondition \( R \). Of course, there may be many \( x \)'s satisfying \( R \). And here we arrive at what we think is a distinguishing feature of the binary search, namely that

it is beyond our control
which \( x \) satisfying the
postcondition will be
generated.

(This is in sharp contrast to linear searches where a fully specified value is computed.)

In order to substantiate this feature we consider the instantiation \( a, b := 0, 100 \) and \( Z := \text{true} \), meeting Prog0's precondition and \( (0) \). We then obtain
\[ x, y := 0, 100 \]
\[ \text{do } y \neq x + 1 \rightarrow h := (x+y) \div 2 \]
\[ \text{if } \text{true } \rightarrow x := h \]
\[ \text{false } \rightarrow y := h \]
\[ \text{od} \]
\[ \{ 0 \leq x \land x < 100 \} , \]

which is a highly nondeterministic program. And the reader may check, in whatever way he likes, that indeed, it can generate any value \( x \) such that \( 0 \leq x \land x < 100 \).

We will return to this distinguishing feature at a later stage.

\[ \star \times \star \]

Meanwhile, some readers may have been puzzled by the "weird" condition (0). But it is not weird at all: when taking the contrapositive of (0),
we get
\[ (0') \quad x \in \neg Z \land h \in (\neg Z) y \Rightarrow x \in (\neg Z y) , \]
and this just expresses that \( Z \)'s complement relation \( \neg Z \) is transitive.

Here are some of such relations \( Z \) (for array \( f \) of the appropriate type):

\[ x \in Z y \equiv \]
\[ \bullet f.x \neq f.y \]
\[ \bullet f.x < f.y \]
\[ \bullet f.x \leq A \land A < f.y \]
\[ \bullet f.x \cdot f.y \leq 0 \]
\[ \bullet f.x \land f.y \]
\* \* \*

The last example in the above list — \( \neg Q.x \wedge Q.y \) — is a frequently occurring one, and we shall now deal in more detail with its probably best-known instance, viz. \( Q.i \equiv C \leq f.i \) . That is, given

- integer array \( f[a..b] \) with \( a < b \)
- integer constant \( C \) satisfying
  \( f.a \leq C \wedge C < f.b \)

our binary search becomes — see remark below —

\[
\begin{align*}
\{a < b\} & \{f.a \leq C \wedge C < f.b\} \\
x, y & := a, b \\
\{\text{inv } P0: a \leq x \wedge x < y \wedge y \leq b \} & \{\text{bnd } y - x\} \\
\text{: do } y \neq x + 1 \rightarrow & \\
\quad h & := (x + y) \div 2 \quad \{a \leq h < y \leq b\} \\
\quad \text{if } f.h \leq C \rightarrow x := h \\
\quad \text{if } C < f.h \rightarrow y := h \\
\text{: od} \end{align*}
\]

\[\{a \leq x \wedge x < b\} \{f.x \leq C \wedge C < f.(x+1)\}\]

**Prog 1**
Remark. Actually, we arrived at the above program not by instantiating our original program scheme Prog0, but by just re-deriving it for the current relation \( \forall x \in C \land C \subset f.y \). That derivation is so simple that it can be done by heart: guided by the invariants \( P_0 \) and \( P_1 \), the program can be written down at once.

End of Remark.

Note that the precondition of Prog0 (and hence of Prog1) is the only property of \( f \) used in the derivation. Thus it should be clear, we hope, that the binary search has a right of existence outside the realm of sorted arrays. There are, however, a few little theorems that do link the algorithm with "sortedness" of array \( f \). We present them as exercises to the reader, but not without pointing out that they admit nice - calculational - proofs.

Exercises

(a) If Prog1's precondition is strengthened with the assertion that \( f[a..b] \) is ascending, then there is only one \( x \) satisfying the postcondition.

(b) If there is only one \( x \) satisfying the postcondition, then we have for that \( x \):

\[
\langle \forall i : a \leq i \leq x \land f.i \leq C \rangle \land \langle \forall i : x + 1 \leq i \leq b : C < f.i \rangle
\]

(This expresses a kind of "sortedness" of \( f \) in that "to the left" of \( x \) all elements of \( f \) are at most \( C \), and "to the right" of \( x \) all elements exceed \( C \).

End of Exercises.
Why is it that the binary search, and in particular Prog1, has been so tightly coupled with sortedness of array $f$? The reason, we believe, is that for an ascending array $f$, the post-condition $a \leq x < b \land f(x) < C \land C < f(x+1)$ of Prog1 is so extremely helpful in recording the presence or absence of value $C$ in array $f[a..b]$

- if $f(x) = C$, the answer is yes, since $a \leq x < b$ — no ascendingness of $f$ is needed —
- if $f(x) \neq C$, we have $f(x) < C \land C < f(x+1)$, and now ascendingness of $f$ implies that all elements of $f$ differ from $C$.

In summary, for ascending array $f$, Prog1 postfixed with the single assignment

$$\text{present} := (f(x) = C)$$

records the presence of $C$ in $f[a..b]$, provided —--yes--, still provided array $f$ and constant $C$ satisfy the rather ad-hoc precondition

$$(i) \quad a < b \land f[a..C] < C < f[b]$$

* * *

We have derived Prog0 (and Prog1) under a precondition — $a < b$ — $a \leq b$. In many practical situations, however, this condition need not be satisfied. There are two general ways out (which, unfortunately, don’t always work). We shall illustrate them for the above example of
recording the presence of value $C$ in an arbitrary ascending array $f[a..b]$, i.e. in an array for which (1) may or may not hold.

The first way out is to try to solve the problem separately for the cases in which the precondition does not hold. For our current example this gives rise to, for instance, the following program:

\[
\text{true}
\begin{cases}
\text{if } b < a \rightarrow \{ \text{array } f[a..b] \text{ is empty} \} \\
\quad \text{present} := \text{false}
\end{cases}
\begin{cases}
\text{if } a \leq b \rightarrow \{ \text{array } f[a..b] \text{ is non-empty} \} \\
\quad \text{if } C < f.a \rightarrow \{ \text{f is ascending, hence} \} \\
\quad \quad \text{present} := \text{false}
\end{cases}
\begin{cases}
\quad f.b \leq C \rightarrow \{ \text{f is ascending, hence} \} \\
\quad \quad \text{present} := (f.b = C)
\end{cases}
\begin{cases}
\quad f.a \leq C \land C < f.b \rightarrow \{ a \neq b, \text{hence } a < b \}
\end{cases}
\text{Prog1}
\begin{cases}
\quad \text{present} := (f.x = C)
\end{cases}
\text{fi}
\text{fi}
\text{fi}.
\]

The case analysis that one is thus forced to introduce is not really an appealing one.

The more elegant way out is based on the observation that Prog1 does not inspect the array elements $f.a$ or $f.b$ at all -vide precondition $a < h \land h < b$ to the inspection of $f.h$ -. This implies that their actual values are completely irrelevant to the (outcome of the) computation:
they are thought variables mainly.

We exploit this observation in the following way. Suppose all we know about \( f[a..b] \) is that it is ascending and that \( a \leq b+1 \) (the array may be empty). We define thought values \( f(a-1) \) and \( f(b+1) \) in such a way that

- \( f[a-1..b+1] \) is ascending, and
- \( f(a-1) \leq C \land C < f(b+1) \), and

since in addition \( a-1 < b+1 \) holds, we have thus laid down precisely the required precondition of \( \text{Progr1} \), be it with the instantiation \( a, b := a-1, b+1 \). The solution to our problem now reads

\[
\begin{align*}
x, y & := a-1, b+1 \\
\textbf{do} & \ y \neq x+1 \rightarrow \\
& \quad h := (x+y) \div 2 \\
& \quad \{ a \leq h \land h \leq b, \text{ hence } f.h \text{ is defined} \} \\
& \quad \text{if } f.h \leq C \rightarrow x := h \\
& \quad \quad \{ C < f.h \rightarrow y := h \} \\
& \quad \text{fi} \\
\textbf{od} & \quad \{ a-1 \leq x \land x < b+1 \} \quad \{ f.x \leq C \land C < f(x+1) \} \\
\textbf{if} & \quad a-1 = x \rightarrow \{ C < f.a \} \text{ present := false} \\
& \quad \{ a-1 < x \rightarrow \{ a \leq x \land x \leq b, \text{ hence } f.x \text{ is defined} \} \\
& \quad \quad \text{ present := (f.x = C) } \\
\quad \text{fi}
\end{align*}
\]

And this, we think, is a much nicer development — and a much nicer code — for the standard binary
search example. The technique applied—adding thought values so as to establish a required precondition—is suitable in many other examples as well.

* * *

We conclude our treatment of the binary search technique with two exercises and a question. The exercises are

(i) Given that integer array \( f[a..b], a < b, \) is the concatenation of an increasing and a decreasing sequence, find the top (the maximum). More precisely, design a program that computes a value \( M \) such that
   \[ a \leq M < M + b \]
   \[ f[a..M] \text{ is increasing} \]
   \[ f[M..b] \text{ is decreasing} \]

(Note: \( f[a..M] \text{ is increasing} \)

\[ \equiv (\forall i: a \leq i < M: f[i] < f(i+1)) \]

(ii) Given that integer array \( f[a..b], a < b, \)
   - is the concatenation of two increasing sequences
   - contains a dip, i.e.
     \[ (\exists i: a \leq i < b: f[i] > f(i + 1)) \]
   - satisfies \( f[a] > f[b] \),

   design a program to compute a dip.

Both exercises can be solved using a binary search.
If, however, in exercise (ii) the given \( f.a > f.b \) is dropped, we no longer know of a way to solve this problem with the binary search technique, not even if we introduce a thought value \( f.(a-1) \) such that \( f.(a-1) > f.b \). The reason why the introduction of such a thought value does not help here is that it may introduce a second dip at \( a-1 \), and, as we said before, it is now beyond our control which dip will be computed by the binary search: it could be the required one, but it could also be the thought dip that we ourselves introduced.

The as yet unsolved problem that remains is to identify the precise conditions for a binary search to be applicable.

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