

Optimization
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Lecture - 27
Fibonacci Method

Today's topic is Fibonacci method. This method is again another elimination technique, for solving single dimensional one variable non-linear optimization problem. The basic necessity for applying this method is that, the function must be unimodal in the initial interval of uncertainty. Now, one thing must be said that this Fibonacci method, the beauty of this method is that it makes use the Fibonacci numbers in the sequence. That is, if there are certain limitations of this method as well. Now, let me just first tell what are the basic necessities and what are the limitations for the Fibonacci method.

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Fibonacci Method.

- This method is an elimination technique.
- Basic necessity is that f must be unimodal.
- It makes use of Fibonacci numbers.

$\{F_n\} \quad \{1, 1, 2, 3, 5, 8, 13, 21, 34, 55, \dots\}$

$F_0 = 1, F_1 = 1$
 $F_n = F_{n-1} + F_{n-2}$

- Total number of experiments

$f(x)$

x^*

$a \quad b$

$I_0 = [a, b]$

Min $f(x)$
subject to
 $a \leq x \leq b$

x^*

$a \quad b$

First of all, this method is an elimination technique. Elimination technique, that is another name can be said as the interval reduction method, for solving for getting optimal for non-linear optimization problem. And the basic necessity for applying this method is that, as I said this is the basic necessity is that the function must be unimodal. Function means the function which I am going to optimize, must be unimodal in the given range.

And it makes use of Fibonacci numbers. As we know, there is a well known sequence of Fibonacci numbers or Fibonacci people as saying in that name as well. If we consider F_n is the sequence, that is the Fibonacci numbers and we must be knowing that this is the sequence we are getting 1, 1, 2, 3, 5, 8, 13, 21, 34, 55 in this way. If we just look at the sequence, we see it follows certain rules. The rule is that, if I consider F_0 is the first number, F_1 is the second, F_2 is the third in that way and we see that F_0 is all over is equal to 1, F_1 is equal to 1. And if we consider other number, that would be sum of the previous 2 numbers. That is why F_n must be equal to F_{n-1} plus F_{n-2} . That is the beauty of the Fibonacci number and we will use these numbers in the Fibonacci method.

And there is one good thing for this method is that, we can we have to define, we have to say the total number of experiments before initiation. Total number of experiments or the total number of approximation we are making for optimal solution, that has to be said beforehand. And if this is so, now let us say this is my, this is my function, this is the given interval from a to b and this is the function.

As I said, function must be unimodal, say we want to minimize the function, minimize $f(x)$ and this is the range x . Then we are having the function minimize $f(x)$ subject to x in between a to b . If we look at the graph, it is very clear that it has the minimum at this point. Let us say, x^* is the minimum for this function. Now, let me see another function. Again also from a to b and this is the function for us. Now we see that function is having minimum at this point, this is my x^* , this is the optimal solution for this function.

Now, we are just we are want to implement one iterative processes that is the Fibonacci method for solving this minimization of $f(x)$. If we just look at the property of the function, we see that function is having minimum here. That is why, we can just have the reduction, interval reduction process in this way that, in the first iteration we will consider the initial interval. Say, initial interval is a to b and we will say this initial interval of uncertainty as L_0 . Now, since we are having minimum here, in the next level we can, we may just discard this part of the interval. So that in the next iteration, we will just consider this interval and in this case, we will just discard may be we may discard this part.

Now, that is why the basic philosophy of the region elimination technique is that similar to Fibonacci method is the as well, in every iteration we reduce a part of the interval, so that we will just we will conclude, we will just complete, we will just conclude our iterative process with a shorter interval, with a small interval where we will declare the optimal is laying in that interval. That is why the process is like that, we will start from initial interval of uncertainty L naught is equal to a to b .

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Given the initial interval of uncertainty $L_0 = [a, b]$ and let the number of experiments be n .

Step 2 To determine x_1 & x_2
define $L_2^* = \frac{F_{n-2}}{F_n} L_0$

$$x_1 = a + L_2^* = a + \frac{F_{n-2}}{F_n} L_0$$

$$x_2 = b - L_2^* = b - \frac{F_{n-2}}{F_n} L_0 = b - \frac{F_{n-2}}{F_n} (b - a) = \left(1 - \frac{F_{n-2}}{F_n}\right)b + \frac{F_{n-2}}{F_n} a$$

$$= \left(\frac{F_n - F_{n-2}}{F_n}\right)b + \left(\frac{F_n - F_{n-1}}{F_n}\right)a$$

$$= \frac{F_{n-1}}{F_n} b + a - \frac{F_{n-1}}{F_n} a$$

$$= \frac{F_{n-1}}{F_n} (b - a) + a$$

$$= a + \frac{F_{n-1}}{F_n} L_0$$

Diagrams:
1. Initial interval $[a, b]$ with points x_1 and x_2 marked.
2. Interval $[a, x_2]$ with points x_1 and x_2 marked.

Now, my next process. My step one that is why is that, for Fibonacci method is that given the initial interval of uncertainty, let me just elaborate the iterative process L naught from a to b . Now as I said, this is the number of experiments must be said beforehand. Let we considered, let the number of experiments be n . Once we are specifying the number of experiments, then since we are using the Fibonacci numbers, that is why we are sure that the we will considered the Fibonacci numbers up to F_n . If we are having 5 number of experiments, then we will use the Fibonacci numbers F_0, F_1, F_2, F_3, F_4 and F_5 in that way, just I will show the thing with a example in the next. Now just, let us go to the next step, that is step 2. What is my basic?

Basic idea is that, we will reduce a part of the given interval of uncertainty, that is a to b . That is why the process tells us that, we will considered 2 points, 2 experiments here. Say x_1 and another one is x_2 , in such a way that there is a specific distance of x_1 and x_2 from both the end of the given interval. In the next step, in the step 2, I will just show

how to get the next approximations for the optimal solutions x_1 and x_2 . That is why, to determine x_1 and x_2 , what we have to do. That is initial 2 experiments, approximations will be defined in this way. We need to define one number, that is L_2^* . The L_2^* would be is equal to F_{n-2} by F_n into L_0 .

Here n is the number of experiments as we have specified before. That is why if it is n is equal to 5, then L_2^* must be is equal to F_3 divided by F_5 into length of the initial interval of uncertainty, that is L_0 . And why we need L_2^* because what we will do, we will consider the interval a to b and we will just select x_1 and x_2 in such a way that, both the experiments x_1 and x_2 these are all L_2^* distance apart from both the ends of the given interval. That is the L_2^* quantity. That is why once our L_2^* is defined in this way, then we can generate 2 experiments, rather 2 approximations for the optimal solution x_1 and x_2 .

Therefore x_1 would be is equal to, since it is L_2^* distance apart from a x_1 must be is equal to $a + L_2^*$. In other way, we can write $a + L_2^*$ is equal to $a + F_{n-2}$ divided by F_n L_0 . What about x_2 ? x_2 is again L_2^* distance apart from b , that is why x_2 must be is equal to $b - L_2^*$. That is equal to $b - F_{n-2}$ divided by F_n L_0 . This is the basic thing, but we can do certain simplification for x_2 here. How we can do so? We know L_0 is equal to that is the length of the initial interval of uncertainty. That is why we can write L_0 is equal to $b - a$.

Once this is so, then we can write it down as $1 - F_{n-2}$ divided by F_n into b plus F_{n-2} by F_n a . This again is equal to $F_n - F_{n-2}$ divided by F_n into b plus F_{n-2} can be written as we know F_n is equal to F_{n-1} plus F_{n-2} . That is the n th Fibonacci number is the sum 2 previous numbers. That is why $F_n - F_{n-2}$ can be written as, $F_n - F_{n-1}$. That is why, we can write it down in this way.

Once this is so, we can write it as $F_n - F_{n-1}$ again F_{n-1} divided by F_n into b plus a because F_n by F_n is 1. This is equal to F_{n-1} by F_n into a . Thus we are getting it as $a + F_{n-1}$ divided by F_n . Am sorry, a will not be there. F_{n-1} into F_n into $b - a$ plus a . Thus we are getting $a + F_{n-1}$ by F_n L_0 . That is why just look at the another form of x_1 , x_2 . This is my x_1 , $a + F_{n-1}$ by F_n L_0 and what about x_2 ? $a + F_{n-1}$ by F_n L_0 . In other way, we can

say this is as $b - F_{n-2} \text{ by } F_n L_0$. This is this is the, these are the forms of x_1 and x_2 respectively. How to adopt in the example?

I will just do in the next. Now if this is so, what is my next task? If the function is like this, if this is my function, then certainly at point x_1 the functional value is lesser than at point x_2 . If we consider the other way, this is a this is b and there are 2 approximations where x_1 and x_2 are lying between a to b . The function is like this. Certainly $f(x_1)$ value, functional value at x_1 must be lesser than functional value at x_2 . That is why, in both the cases if we consider the elimination technique, that is very clear in the case one, we will reduce this interval in the next iteration because minimum cannot lie here.

And if we consider, the next case, case two: then we must reduce this interval from x to b . That is why, if we just consider both the things together, what is my next task? My next task is to find the values $f(x_1)$ and $f(x_2)$ and by considering the unimodality property, we will reduce either a to x_1 , then my new interval of uncertainty would be x_1 to b or in the next case my new interval of uncertainty would be from a to x_2 . This is the case for both the cases.

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Step 3 L_2 would be either $[x_1, b]$ or $[a, x_2]$

$$L_2 = L_0 - L_2^* = L_0 - \frac{F_{n-2}}{F_n} L_0$$

$$= \left(1 - \frac{F_{n-2}}{F_n}\right) L_0 = \frac{F_{n-1}}{F_n} L_0$$

Step 4 To evaluate x_3

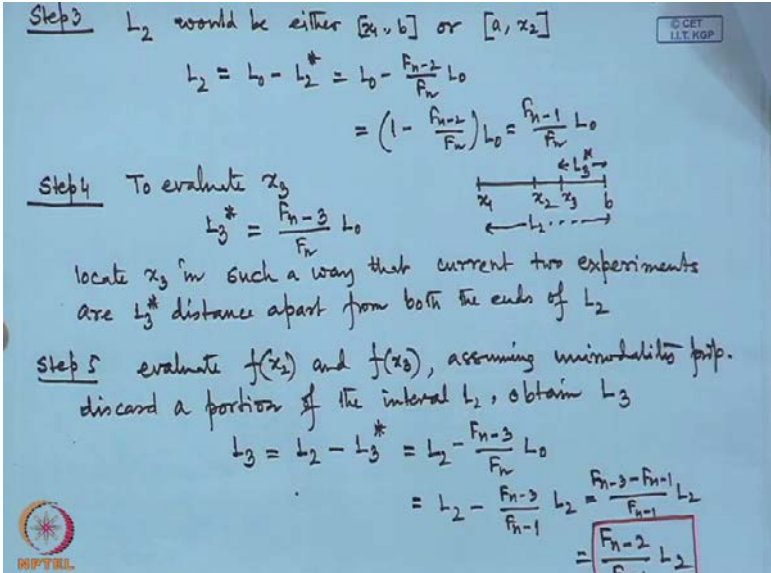
$$L_3^* = \frac{F_{n-3}}{F_n} L_0$$

locate x_3 in such a way that current two experiments are L_3^* distance apart from both the ends of L_2

Step 5 evaluate $f(x_2)$ and $f(x_3)$, assuming unimodality prop. discard a portion of the interval L_2 , obtain L_3

$$L_3 = L_2 - L_3^* = L_2 - \frac{F_{n-3}}{F_n} L_0$$

$$= L_2 - \frac{F_{n-3}}{F_{n-1}} L_2 = \frac{F_{n-2} - F_{n-1}}{F_{n-1}} L_2$$

$$= \frac{F_{n-2}}{F_{n-1}} L_2$$


Now, that is why my next step would be, step three would , we will reduce the interval L naught and we will rename the new interval as L_2 as a new interval of uncertainty. That is why my new interval of uncertainty L_2 , that would be either as we see from the

previous case, either it is x_1 to b or it would be from a to x_2 . Accounting to the functional value at x_1 and x_2 .

Now, what is the length of L_2 then? What is the next interval of uncertainty let length? Then certainly it would be from x_1 to b or a to x_2 and we will see the nice fact that L_2 will be is equal to $L_{\text{naught}} - L_2^*$ because we are reducing this length. That is L_2^* length, this is length we are we have considered x_2 , x_2 from b L_2^* distance apart. That is why next interval of uncertainty that would be is equal to L_2 and this L_2 would be is equal to $L_{\text{naught}} - L_2^*$. What is L_2^* again?

Let me write it down, L_2^* that would be $F_n - 2$ by $F_n L_{\text{naught}}$ and that can be written very nicely as $F_n - 2$ by $F_n L_{\text{naught}}$ and the nice fact is that L_2 would be is equal to $F_n - 1$ by $F_n L_{\text{naught}}$. That is my next interval of uncertainty. Now once it is fixed, then we will just find out the next approximation. In the next step, step 4. That is why we will go for next approximation. If it is from x_1 to b , we are having x_2 here. Then we will go to evaluate the next approximation x_3 in the next.

Now to define x_3 , again we will have to evaluate L_3^* . L_3^* would be is equal to, as we have seen $F L_2^*$ is equal to $F_n - 2$ by $F_n L_{\text{naught}}$, L_3^* would be is equal to $F_n - 3$ by $F_n L_{\text{naught}}$. Again, locate x_3 in such in a way that, current 2 experiments are L_2 L_3^* distance L_3^* distance are apart from both the ends of the interval of uncertainty. Current 2 experiments are L_3^* distance apart from both the ends of L_2 . This is my L_2 . That is why it would be from that must be position of x_3 in the next case. And if this is so, then again the same process, we will just repeat the process like previous. We will go the step 5. We will evaluate the value, functional value at x_2 and f at x_3 and assuming the unimodality property, we will discard a part of the interval.

Unimodality property, discard a portion of the interval L_2 and obtain the new interval of uncertainty L_3 . And what would be the length of L_3 ? Certainly, since we are discarding this is L_3^* , then the new interval of uncertainty would be L_3 would be is equal to $L_2 - L_3^*$. Again if we just write down the thing in detail, that must be is equal to $F_n - 3$ by $F_n L_{\text{naught}}$ and this is equal to $F_n - 3$, that can be written as $F_n - 1$, L_2 as well. It would be must be is equal to $F_n - 3$ minus $F_n - 1$ divided by $F_n - 1 L_2$ and this would be is equal to $F_n - 2$ divided by $F_n - 1 L_2$. And this is the form of L_3 .

If you look at the pattern of L 2 and L 3 separately, in L 2 we got $F_n - 1$ by $F_n L$ naught in L 3 we are getting $F_n - 2$ by $F_n - 1 L 2$.

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For obtaining the j^{th} experiment

$$L_j^* = \frac{F_{n-j}}{F_{n-(j-2)}} L_{j-1}$$

Length of interval of uncertainty after j^{th} experiment

$$L_j = \frac{F_{n-(j-1)}}{F_n} L_0$$

Measure of efficiency = $\frac{L_n}{L}$

That is why in general, for obtaining the n th experiment. Because as we got x 3, in the similar process we will discard a part of L 3 and we will reach to L 4 and there also we will get another experiment x 5. In that way, we will proceed further and further. How long we will be proceed? We will proceed up to the n th experiment, That is why, if we consider the j th experiment, the formula would be, to obtain the j th experiment point, we need to find out L_j star as we got L 2 star and L 3 star and that must be is equal to $F_n - j$ divided by $F_n - j - 2 L_j - 1$.

And the corresponding length of uncertainty after the j th experiment, L_j would be is equal to, that is the length of the interval of uncertainty would be $F_n - j - 1$ divided by $F_n L_0$. If I put L_j is equal to 2, we are getting $F_n - 1$ divided by $F_n L_0$, if j is equal to 3, then we are getting $F_n - 2$ divided by $F_n L_0$. In that way, we can generator all the points x 1, x 2, x 3 up to x n and we will reduce the interval further and further and at end we will get n th interval of uncertainty and the middle of that interval of uncertainty will be declared as the optimal solution for the given problem. Whatever we I have said, let me just let me just apply the whole process in the in one of the example.

But one thing I must tell here, as I have said for other elimination technique that there is one measure, that is called the measure of efficiency, efficiency of the elimination technique because we have learnt the exhaustive search technique dichotomous search, interval search technique and this is another searching technique that is the Fibonacci method. In the next, we will do the golden section method. For every method, there is a there is a measure for a efficiency. In other another name that is called the reduction ratio, through which we can judge the efficiency of the of the elimination process. And I will discuss more on this, but here just I want to tell you the measure of efficiency or the reduction ratio can be defined as $L_n - L_0$, where L_n is the length of the interval of uncertainty after n th experiment and L_0 is the initial interval of uncertainty.

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Ex 1 Find the minimum value of $f(x) = x^2 + 2x$ within the interval $[-3, 4]$ using Fibonacci method. Obtain the optimal value within 5% of exact value.

Length of final interval of uncertainty $\leq \frac{5}{100} \times \text{Length of initial interval of uncertainty}$

$\Rightarrow \frac{L_n}{2} \leq \frac{1}{20} L_0$

$\Rightarrow L_n \leq \frac{L_0}{10}$

$\Rightarrow \frac{L_n}{L_0} = \frac{1}{F_n} \leq \frac{1}{10}$

$\Rightarrow F_n \geq 10$

$n \geq 6$

$F_0 = 1$
 $F_1 = 1$
 $F_2 = 2$
 $F_3 = 3$
 $F_4 = 5$
 $F_5 = 8$
 $F_6 = 13$
 $F_7 = 21$

Now this is the thing. Now, let me apply this process for one example. A example is that, find the minimum value of $f(x)$ is equal to $x^2 + 2x$, within the interval minus 3 to 4 using the Fibonacci method and obtain the value, optimal value within 5 percent of exact value. Now, this is the problem for us now we want to minimize the function $x^2 + 2x$, within the interval of minus 3 to 4. If we just draw the function, we will see the function is very much unimodal, within minus 3 to 4 and function is having a minimum somewhere in between minus 3 to 4 and there is one restriction is that we want to have the optimal value within 5 percent of exact value. We need to use the Fibonacci method.

As I said the first necessity function must be unimodal, that is true here the second is that the number of experiments must be specified beforehand. But there is no information as such about the number of experiment. But we can deduce it from the given desire accuracy, the we want to have the accuracy 5 percent accuracy of the exact value, 5 percent of the exact value. That is why if we say L_n is the interval after n th experiment, if this if this is my L_n , then mid value will be declared as the optimal value I said. Then if you want to have the 5 percent of the exact value, then the half of the n th interval of uncertainty must be less than is equal to L_0 into error percentage. That is why we can write it down that, what is my objective?

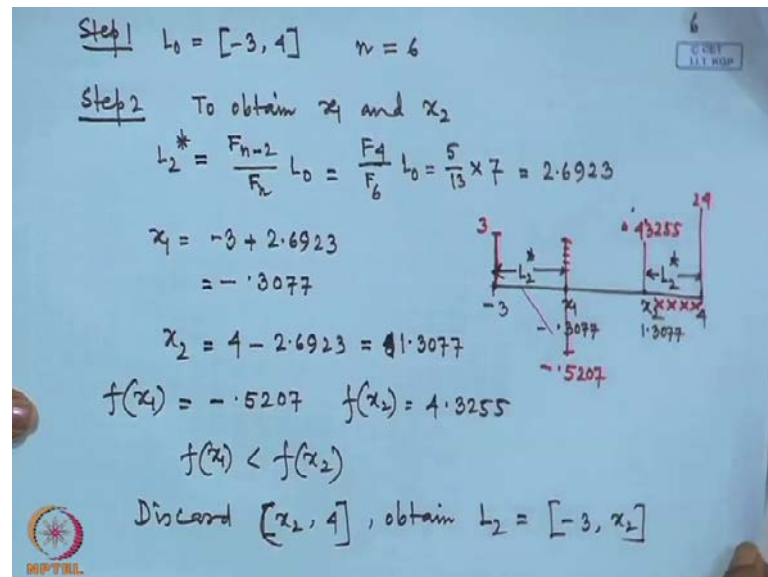
My objective is to find out the number of experiments from the given, from the level of desire accuracy, length of final interval of uncertainty after n th experiment, divided by length of initial interval of uncertainty. Must be less than is equal to 5 percent. That means, actually we are considering half length of the interval of this is must be is equal to 2 into this. That is why, this must be is equal to 5 by 100. In other way, we can say as L_n by 2 must be less than is equal to 1 by 20 L_0 or we can say L_n must be is equal to less than is equal to L_0 by 10. One nice fact of Fibonacci method is that, as I said the measure of efficiency L_n by L_0 , that is after the n th experiment, if L_n is the n th L_n is the interval of uncertain, L_0 is the initial interval of uncertainty, always we will see for any number of n . Whatever number of n we will choose, we will see L_n by L_0 always will be is equal to 1 by F_n . If we consider n is equal to 5, it would be 1 by F_5 . If we consider n is equal to 6, it would be 1 by F_6 .

That is why other way, we can say that this measure this reduction ratio can be used for selection of the number of experiments, if it is not given to us. That is why, we can say that L_n by L_0 that is equal to 1 by F_n must be less than is equal to 1 by 10. What exactly we are getting, F_n must be is greater than equal to 10 and we know F_0 is equal to 1, F_1 is equal to 1, F_2 is equal to 2, F_3 is equal to 3, F_4 is equal to 5, F_5 is equal to 8, F_6 is equal to 13. That is why we can conclude here that since F_n is greater than equal to 10, that is why we can say n must be greater than equal to 6. This is my conclusion, that is why from here we can say the minimum 6 number of experiments we have to do for getting the desired level of accuracy.

That is the 5 percent of exact value. As we have considered the middle point of the middle point of L_n as the optimal solution of the of the given problem. That is why we

are concluding from here that the minimum number of n must be 6. If we just increase the number of n as 7, 8 etcetera then we will get better and better result. That is why the interval of uncertainty will be smaller and smaller further. But since, we are doing it manually that is why we will restrict ourselves to value of n as equal to 6.

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We will apply in the next the iteration process from step 1. What is my step 1, initial interval of uncertainty is given for us. That is why step 1 would be is equal to L_0 would be equal to from minus 3 to 4 and my n is given we are considering n is equal to 6.

Now, step 2. We have to obtain 2 approximations x_1 and x_2 . Now for that thing, as I said in the as I said in the algorithm, we have to generate L_2^* . how to get L_2^* ? L_2^* star must be is equal to F_{n-2} divided by F_n L_0 . What is my F_{n-2} here? My n is equal to 6. That is why it must be is equal to F_6 by F_4 by $\times 6$ L_0 . What is F_4 ? F_4 equal to 5. As we have seen F_4 equal to 5, 5 by 13 into L_0 , the length. Here length is coming as 7. That is why we will multiple with 7 and this value we will 2.6923. Once we are getting L_2^* , we will generate 2 experiments x_1 and x_2 in the next. How to get x_1 ?

As I said, if I just consider the interval a to b , that is from minus 3 to 4, then x_1 and x_2 we will generate in such a way that both the experiments will be L_2^* distance apart from both the n 's. That is why x_1 must be is equal to minus 3 plus L_2^* , that is 2.6923 and we are getting x_1 is equal to minus 0.3077. And if we consider x_2 , we will

get 4 minus 2.6923 and that would be is equal to 4 1.3077. That is 1.3077 and function is unimodal in between, that is why our next task is to find out the functional values at x_1 and x_2 . What is the functional value at x_1 ?

Let us see, before to that let me see the functional value at both the ends. If we consider here the functional value with will come as 3 and $f(x)$ will be 24 at 0.4. At x_1 , we will see the functional value is coming, $f(x_1)$ is coming as, that is minus not in the positive side. That is in the negative side minus 0.5207 and x_2 value is coming $f(x_2)$ value, that is coming 4 3 2, 4 point sorry 4.3255. If we just write down the functional values here, $f(x_1)$ would be is equal to, we know the function is x square plus $2x$ if we just substitute the x_1 value there, we will get $f(x_1)$ that would be minus 5207 and $f(x_2)$ is coming as 4.3255. What we see that, $f(x_1)$ is lesser than $f(x_2)$.

Since the function is unimodal, the maximum I am sorry, this is the minimization of the if function. The minimum cannot live within this because the functional value function is coming like these only. That is why minimum will be in the within the interval from minus 3 to x_2 . That is why we are discarding. This is the conclusion that discard x_2 , 4 and obtain the new interval of uncertainty, obtain L_2 that is the new interval uncertainty from minus 3 to x_2 .

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Step 3 $L_2 = [-3, x_2] = [-3, 1.3077]$

$L_2 = L_0 - L_2^* = 7 - 2.6923 = 4.3077$

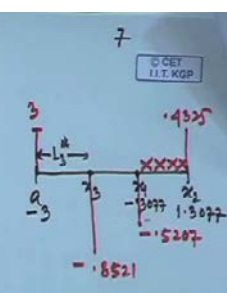
$\frac{F_{n-1}}{F_n} L_0 = \frac{F_5}{F_6} \cdot 7 = \frac{8}{13} \times 7 = 4.3077$

Step 4 To obtain x_3

$L_3^* = \frac{F_{n-3}}{F_n} L_0 = \frac{F_3}{F_6} \times 7 = \frac{3}{13} \times 7 = 1.6154$

$x_3 = -3 + 1.6154 = -1.3846$ $f(x_3) = -1.8521$

$f(x_3) < f(x_1) \Rightarrow$ discard (x_1, x_2)



Coming to the next, step 3. We will get L_2 as we have considered L_2 is equal to minus 3 to x_2 . That is means minus 3 to 1.3077. If we consider, if we just apply the formula we have learnt, just now learnt we will get the same value for L_2 . How?

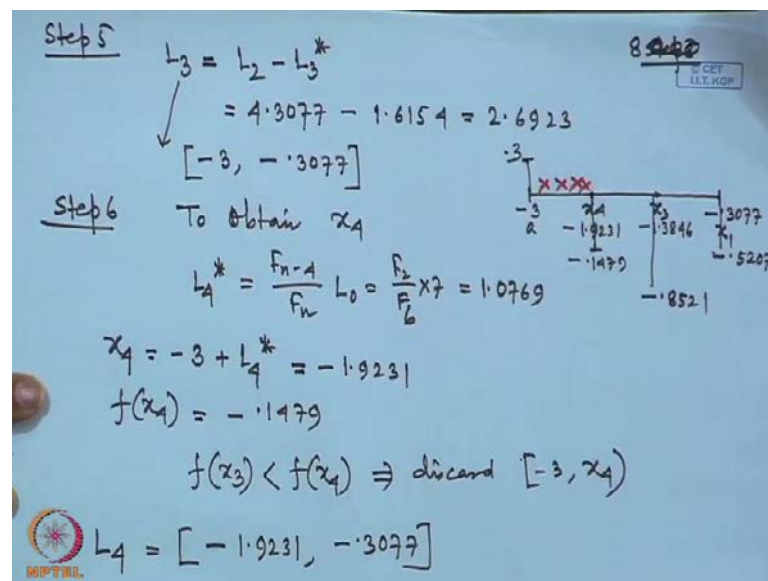
As we know, L_2 is equal to L_{naught} minus L_2^* . What is L_{naught} , that the length of L_{naught} would be 7. Whenever I am writing L_{naught} L_2^* within the equation it means that, this is the length of the given interval. And L_{naught} is 7 and what about L_2^* . Just now we got L_2^* is equal to 2.6923 and that is coming as 4.3077. If we just see the interval size here, it is coming 4.3077 and again there is another nice fact is that, we can get L_2 in this way as well. As we got that L_2 is equal to F_n minus 1 by $F_n L_0$. What is F_n minus 1? That is my F_6 , that is my F_5 . L_0 is 7 this is another formula for L_2 . This is coming as 8 by 13 into 7 is equal to 4.3077. That is the nice fact, this is the beauty of the Fibonacci method.

In if we apply this, if we just consider this one, its coming the same value here also the same value because we are considering here the Fibonacci numbers. But the basic idea is that now in the next we will consider the interval from a to x_2 . What is my a ? That is minus 3, my x_2 is 1.3077. In between, there is a point we are having that is x_1 . My x_1 is equal to minus 0.3077, that is my x_1 . Then what is my next step? Next step would be is equal to obtain the next approximation for the optimal solution, that is x_3 . For getting x_3 again we need to generate L_3^* . As we have done before, L_3^* is equal to F_n minus 3 divided by $F_n L_0$. What is F_n minus 3? This is F_6 , this is F_3 . L_0 is again 7 here and F_6 is equal to 13 and this F_3 is equal to 3. That is why, 3 by 13 into 7 this value is coming 1.6154.

Once we are getting so, very nicely we can select x_3 in such a way that, the current 2 experiments, that is the experiment 1 is x_1 and another one is x_3 . That would be L_3^* star distance apart from both the ends of the given interval. And if we just see the fact x_1 is L_3^* star distance apart because if I just consider the length of the interval from x_1 to x_2 , always it would be is equal to 1.6154. That is why we have to select x_3 here. That where x_3 would be L_3^* star distance apart from one end. And in this way, we will get x_3 is equal to my a plus L_3^* . That is, minus 3 plus 1.6154 and it will come as minus 1.3846 and if we just find out the functional value at this point, we will get the functional value as minus 0.8521.

Now, let us see what is happening in the figure and if we see again at point minus 3 the value is coming 3. At point x 2 the value is, we calculated the value is 0.4325. At point x 1, the value is coming minus 0.5207. At point x 3, the value is coming minus 0.8521. What we see again? Again we see that, $f(x_3)$ is lesser than $f(x_1)$. That is why what is my conclusion? Discard this interval from optimum cannot lie within this because minimum should lie within this interval. That is why discard the interval x_1 to x_2 and we getting new interval of uncertainty L_3 as minus 3 to minus 0.3077.

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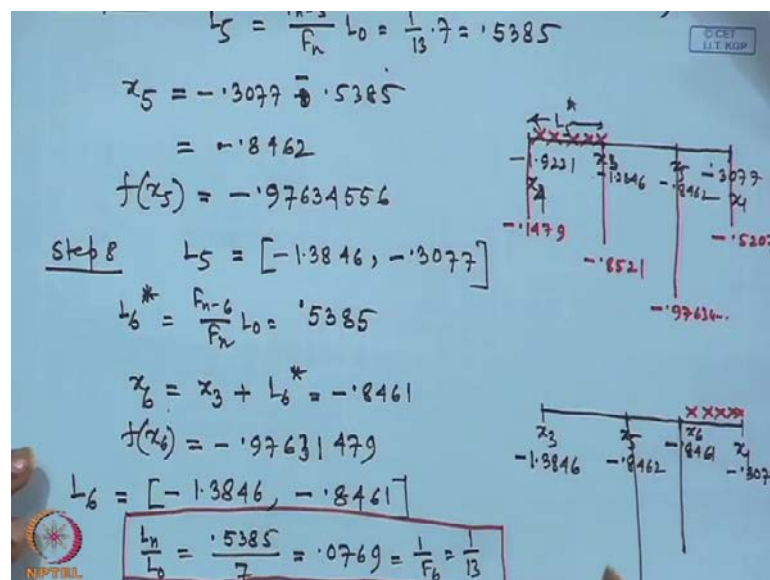
That is why let us go to the next step, step 5. I am sorry, this is not the step 5. This is page numbers 8 and we are going to step 5 and we will get L_3 equal to L_2 minus L_3^* . My L_2 was the length of L_2 rather 4.3077 and L_3^* , just now we have calculated 1.6154. That is why the length is coming 1.6923.

If this is so, then we are reaching to the next interval. This interval, L_3 interval will be from minus 3 to minus 0.3077. What is the next task? Again the next task would be to find out the next approximation. We have to reach up to sixth experiment because we have consider n is equal to 6. That is why we are moving to the next. We will obtain x_4 . Let me draw the figure again. We are having minus 3 in 1 end, minus 0.3077 at x_1 in the other end. This is my a , in between we are having x_3 and this value is, x_3 is equal to just now we got that is minus 1.3846.

Now, to obtain x_4 again, we need to find out L_4 star. That would be is equal to F_n minus 4 divided by $F_n L_0$. Let me just calculate the entire thing very quickly. That would be F_2 into F_6 into 7, that will come as 1.0769. If this is so, again the same thing is that, both the current 2 experiments must be L_4 star distance apart from both the ends. That is why we will consider the next, that is x_4 as minus 3 plus L_4 star and it would be is equal to minus 1.9231 and here must be minus 1.9231. Again we will see the functional value at x_4 . We will see the function, we see the functional value is coming as minus 1 point 0.1479. What we see the figure here again, here it is 3, here it is minus 0.5207.

Here it is minus 0.8521 and here at x_4 the value is coming as minus 0.1479. What we see again? We see that $f(x_3)$ is lesser than $f(x_4)$, then what is the conclusion? Discard the interval from minus 3 to x_4 . Minus 3 to x_4 means we are discarding this interval in the next, that is why the new interval of uncertainty. That is my L_4 will become as from x_4 , that is minus 1.9231 to minus 3077. We have to repeat the process again let us go to the next.

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We have to have more number of experiments, then only we will get better result. We cannot start with n equal to 2 or 3. This is a process, if you just automate the process, the process is very easy. But if we just do it manually, it takes time. L_5 star, L_5 star again

would be equal to $F_n - 5 F_{n-5}$ by $F_n - L_0$, that would be is equal to 1 by 13 into 7 that is coming as 5385.

Therefore, we will again considered the interval from x_3 to x_1 . That was minus 1.9231 to minus 0.3077, that is my x_1 , that is my x_3 here. In between we are having x_4 . Now that is my x_4 . We are having x_3 in between. x_3 is coming somewhere. Now, we will calculate x_5 . Again the same logic, current two experiments, that is x_3 and x_5 must be $L_3 - L_5$ star distance apart from both the ends. If we see that it will be always this is L_5 star. That is why x_5 must be here somewhere, that is why we will consider x_5 is equal to minus 0.3077 plus L_5 star, that is 0.5385. If we consider, we will get x_5 is equal to minus 0.8462. This x_3 was minus 1.3846 and we are having x_5 as minus 0.8462. This is minus.

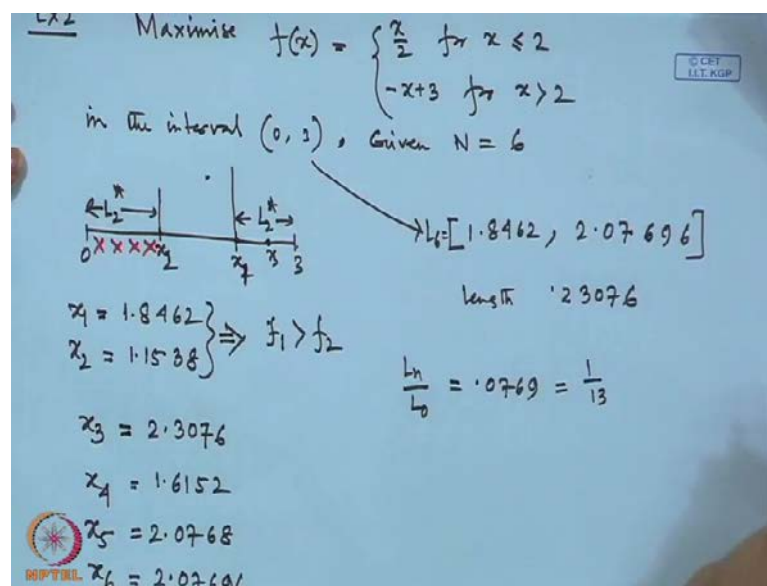
And we will calculate the functional value at x_5 and this value is coming 0.97. We are taking a large number here, with more decimal points because we will see why we are considering in the next. Because once we will go to x_6 , that will be very close to x_5 in the next we will see and that is why, we will see that in the next case functional value will differ with a small amount. That is why we are considering a more decimal points in this case. And here we are having x_4 , this is my x_4 . Now let us see the functional values at every point. At x_4 , it was minus 0.1479 and x_3 , the value was minus 0.8521. At x_5 the functional value is more, that is 0.976344556 and at point x_1 the functional value is minus 0.5207.

Again the same conclusion, since the function is an unimodel x_5 is lesser than x_3 . Therefore, optimal cannot lie in this area. That is why, we will discard this interval and we will get the next interval L_5 as minus 1.3846 to minus 0.3077. Alright and in between how many points? We are having from the previous one, we are having x_3 , we are having x_5 . Let me write down the values here, 1.3846. My x_5 was minus 8462, my x_1 was minus 3077. We need to get x_6 because that is my last approximation. Let us see what is happening in the next. We will consider L_6 star, that would be $F_n - 6$ by $F_n - L$. That is again F_0 by F_6 by into L_0 . That is why, its again 1 by 13 into 7, that is why the L_6 star value is same as the previous value. Always we will see the last, L_n star would be is equal to $L_n - 1$ star always, if the n is number of experiments.

If this is so, then we will say the same thing, that the current 2 experiments would be 6 stars distance apart from both the ends. That is why x_6 would be equal to x_3 plus L_6 star and we will get this value as minus 8461. And what is the functional value at x_6 ? We will see the functional value is coming minus 0.9761. Just see with the $f(x_5)$, 61631 sorry 631479. That is why, what we see we are getting x_6 here, that is minus. This is my x_6 , 8461 and we are having lesser functional value, higher functional value than x_5 . That is why, what is the conclusion in the next? Since this is high value, considering the unimodality condition, again we will discard this part of the interval and we will declare that the final interval of uncertainty L_6 would be equal to from minus 1.3846 to minus 0.8461.

This is the and the middle point of this interval will be declared as the optimal solution, for this interval. Now there are few things to be noted here. The first thing is that, always we will see that, x_{n-1} is almost same as x_n and only the difference will be there in the decimal, higher in the higher decimal points. This is almost same with x_n is same as x_{n-1} . And another nice fact is that, always we will see that L_n by L_0 , that is my reduction ratio that value here, if we see the L_n distance, this distance will be equal to 5385 and L_0 is my 7. This is equal to, we will see 0.0769 and that would be equal to 1 by F_n .

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What is my n here? n equal to 6. That is why it would be 1 by F_6 , rather this will be 1 by 13. That is the nice fact of the Fibonacci method. Always we will get this one. For whatever n value we will consider, is irrespective the value of n , we will get the same result here. Let us move to the next problem and this is our maximization problem.

Let me do the problem very quickly because I will just write down the steps. Without going into detail about the calculations, my problem is that maximize $f(x)$, x by 2 for x lesser is equal to 2. And this is minus x plus 3, for x greater than 2. We must see the pattern of the function. Pattern of the function is very small here. I am sorry, the pattern of the function is that function is not really continuous and we need to maximize within the interval 0 to 3.

And using the Fibonacci method and another thing is given here, that number of experiment has been specified n is equal to 6, alright. For starting the process, again we will do the same process. We will consider a and b , that is 0 to 3. We will consider L_2^* and we will find out x_1 and x_2 . L_2^* distance apart from both the ends. And we will see x_1 will come as 1.8462, x_2 will become x_2 will be 1.1538. We will find out the functional values at both the points here. And we will see the functional value will come, just you calculate. We will see, this will give you the value - f_1 is higher value than F_2 . Since the problem is the maximization problem, that is why the discard process is the reverse to the minimization problem.

Here, we will not discard the higher value. We will discard the lower value. That is why the this would be this one, this is the higher value and we will discard this interval in the first case. And we will move to the next. We will calculate L_3^* and once we will calculate L_3^* , we can generate x_3 . x_3 would be is equal to 2.3076 and this is my interval and we will place x_3 in such a way that, it would be L_3^* distance apart from both the ends. And we will see that x_3 will be, I am sorry this is x_2 and this is coming x_1 . And if we, this is so, we will consider x_3 here and we will get the functional value and functional value will, again we will see that $f(x_3)$ value is lesser than $f(x_1)$ value. In that way, we can complete the whole process.

Let me write down the approximations only. x_1 , x_2 , x_3 , x_4 is coming as 1.6152. x_5 is coming as 2.0768 and x_6 ultimately will come as 2.07696. And if we just do the, apply the same process, this initial interval of uncertainty will reduce at the end to 1×10^{-6} to

2.07696. That would be L_6 , rather because we are doing the 6 number of experiments, all together. Here the length of L_6 is coming.

Again 0.23076 and here also we can just see the same fact, that L_n by L_0 that is the L_6 by L_0 in that case. This divided by 3 is equal to we will get 0769. The same value 1 by 13, 1 by F_6 . The entire calculation, anybody can do just like the previous problem, just only thing just I wanted to mention here that, for the maximization problem, if we just that unimodality property, instead of considering the lower value of function, we will considered the higher value of function in the selection.

And that is why the discard, the elimination of the interval will be according to that and in that way we will reach to the final interval of uncertainty. That would be much more smaller length and that can be declared as the final interval of uncertainty and middle value of that final interval of uncertainty will be the optimal solution. Fibonacci method is very efficient method for getting optimal solution, for non-linear programming problem even for discontinuous function. That is all for today.

Thank you.