

Convex Optimization
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Lecture No. # 31

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General framework of an Interior point algorithm (General IP)

- Starting vector $(x^0, y^0, s^0) \in \bar{F}$
- Accuracy requirement: $\epsilon > 0$

Initialize $(x, y, s) \leftarrow (x^0, y^0, s^0)$, $\tau \leftarrow \frac{1}{n} \langle x, s \rangle$

While $\tau > \epsilon$

Solve
$$\begin{pmatrix} 0 & A^T & I \\ A & 0 & 0 \\ S & 0 & X \end{pmatrix} \begin{pmatrix} \Delta x \\ \Delta y \\ \Delta s \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ -Xs + \sigma e \end{pmatrix} \rightarrow (*)$$

where $\sigma \in [0, 1]$.

$(x, y, s) \leftarrow (x, y, s) + \alpha (\Delta x, \Delta y, \Delta s)$

$\alpha \rightarrow$ scale parameter (step size) $\alpha \in [0, 1]$

$\tau \leftarrow \frac{1}{n} \langle x, s \rangle$ end.

So, before we progress to showing that the IP algorithms are actually polynomial time, it is good that we take a short look at what we had done yesterday. So, we will just go over it very fast. So, this is what we did in the last class was to talk about the general framework of an interior point algorithms, where you have a starting point, and then you initialize the duality measure, and you solve not exactly the relaxed k k t condition. But some little modification of it using the centering parameter sigma, and that would once done; this is the way you update from a current solution to a new iterate. And then we did certain estimates which allowed us to show that when I change from estimate x_k to x_{k+1} , my duality measure is actually going down, because the whole aim of the interior point methods is to pull down the duality measure and force it to 0.

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See if our algorithm is in polynomial time

Complexity result

Given $\epsilon \in (0,1)$ is the accuracy parameter. Suppose we have the following rule

$$\tau_{k+1} \leq \left(1 - \frac{\delta}{n^\omega}\right) \tau_k, \quad k = 0, 1, 2, \dots$$

for some $\delta > \omega$ positive

Also assume $\tau_0 \leq \frac{1}{\epsilon}$ for some $\alpha > 0$

Then there exists an index K with

$$K = O\left(n^\omega \log \frac{1}{\epsilon}\right)$$

Such that $\tau_k \leq \epsilon \quad \forall k \geq K$

$O(f(n)) \leq C g(n)$

$K \leq c n^\omega |\ln \epsilon|$

So, our aim now, is to see, **see** if our algorithm is in polynomial time. Now it would take quite a while, and it is quite technical to go into the details of what it means by polynomial time. Again I would like to stress looking at the wide range of audience with wide range of capabilities and backgrounds. **I would say that** I would like to again assert that polynomial time means, an algorithm which works fine; an algorithm working in polynomial time means, an algorithm is working fine. In the sense that I can tell you that the number of steps required to stop the algorithm is bounded by a polynomial, which is based on the number of data inputs or the say $x \cdot n$ in the $r \cdot n$ plays the quite an important role in the polynomial. So, it is a polynomial in n .

So, **you know that** if I know what is n ? I can say that within this number of steps my algorithm is actually going to finish up. The lot of issues here, but we are not getting into this too much of detail. So, let me start with this complexity result: The statement goes as follows, and this is from these book primal dual algorithms for interior point methods by Stephen J. Wright. It would be good that if you have a look at the book I am showing, **it is** I think this is much better technique than writing it down. You could see the whole title, the publishers Siam and the author

So, let us take epsilon between 0 and 1. This epsilon as I told you before, this is if you look at the frame work, what we intend to achieve? We intend to achieve, as we work on till epsilon is less than equal to epsilon; till tau is strictly bigger than epsilon, we continue

doing this; repeating the courses and τ is less than equal to ϵ we will stop the algorithm.

So, is the accuracy parameter? So, we are driving τ to 0; we need not reach 0, but if we come very near 0, that would **almost all** almost acceptable for us. So, as I want to assert you every time that optimization is not really the science of finding the exact solution, but it is really a some sort of science, rather than mathematics of finding something which is satisfactory and how much satisfactory? That is a question. That **brings into** brings **into** in front of us whole gamut questions that need to be discussed. But for us by viewing a good or a happy thing or something, which I rather, I have been satisfied **with** is that, if τ is less than ϵ ; that is what I want; nothing more. So, that is my requirement.

So, this is my parameter. Suppose we have the following rule, that is you are generating a sequence, where you have the following rule that the duality parameter at the k plus 1 iteration is satisfying this expression; so, this recurrence relation for some δ and ω positive. Now, also assume that when you start the algorithm **μ naught sorry** τ naught, the starting duality measure. So, you have the starting point x naught, y naught, s naught; so, x naught, inner product s naught divided by n .

So, starting duality measure **is this** has satisfies this property, for some may be its greater to give a different notation. **for some I guess** I would say some α or some α greater than 0. So, we were assuming this. We can device α algorithms, which will actually follow this sort of property, and that is what we are going to show next, that we can device an algorithm, where these properties are followed.

So, μ naught is less than equal to 1 by α , for some α s strictly bigger than 0. Then, we are telling, then not μ naught, I am making a mistake τ naught. Then there exist an index k ; k is natural number; so, x of k , that is where you stop; there exist an index k , with k equal to big O of n ω natural log of ϵ . So, you see such that τ of k is less than equal to ϵ for all k , which is bigger than this k . Now, this capital O ; so, capital O of function say f n means this is nothing but a constant times f of n ; it is less than a constant time f of n . This is the meaning of big O . So, these are actually asymptotics, which tells you, **how when how fast** how fast this is moving with respect to this first; how fast this quantity is moving with respect to f n .

So here, it means this is nothing but some constant times, this into this; that is what it means. And you see n to the power ω ; so, ω is some positive number, and n is obviously natural number. So, this is actually a polynomial in m . So, if I know m , and if I fix my accuracy constant, I can put this and I can tell you that within these steps **say some**. So, I can say that, k is actually like this; some constant which is **which** let us for the time being of course, you do not know the constant.

So, k is not more than this. So, within this number of steps at the max, you will reach this conclusion. So, that is what it tries to say. Now we are going to try to prove, what is in here; how do we prove these thing? Now, **our** we will start with our initial expression; the τ_{k+1} is less than $1 - \frac{\delta}{n^\omega} \tau_k$. I would take logarithms on both sides; \ln means natural logarithm log two base e . So, you can do, take the log, because log is in increasing function. Look at **the** those, who have forgotten about this poor function called log. This is the graph of y equal to **log of x** \log of e^x , which is same as $\ln x$.

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$$\tau_{k+1} \leq \left(1 - \frac{\delta}{n^\omega}\right) \tau_k$$

$$\ln \tau_{k+1} \leq \ln \left(1 - \frac{\delta}{n^\omega}\right) + \ln \tau_k$$

$$\ln \tau_k \leq k \ln \left(1 - \frac{\delta}{n^\omega}\right) + \ln \tau_0$$

$$\leq k \ln \left(1 - \frac{\delta}{n^\omega}\right) + \alpha \ln \frac{1}{\epsilon}$$

$\tau_0 \leq \frac{1}{\epsilon^\alpha}$
 $\ln \tau_0 \leq \alpha \ln \frac{1}{\epsilon}$

$\log(1+x) \leq x \quad \text{if } x \geq -1 \rightarrow \text{(Homework!!)}$

$\ln \tau_k \leq k \left(-\frac{\delta}{n^\omega}\right) + \alpha \ln \frac{1}{\epsilon}$

$\tau_k \leq e \quad \text{if } k \left(-\frac{\delta}{n^\omega}\right) + \alpha \ln \frac{1}{\epsilon} \leq \ln e$

$\ln \tau_k \leq \ln e$
 $\tau_k \leq e$

$\frac{\delta}{n^\omega} < 1$

So, here if I take the log, log is increasing function. So, x less than y means, $f(x)$ less than $f(y)$; \log of xy is $\log x$ plus $\log y$, standard result. This is what you have. Now, if you repeatedly apply this formula, $\ln \tau_k$; τ_k is again less than $1 - \frac{\delta}{n^\omega} \tau_{k-1}$; and so you repeatedly apply, so, if you do that you will get a general formula of this form, that $\ln \tau_k$ is less than equal to k times, $\ln \left(1 - \frac{\delta}{n^\omega}\right) + \alpha \ln \frac{1}{\epsilon}$.

$\delta n \omega$ plus $\log \tau_0$. Now, what was τ_0 ? τ_0 is less than 1 by e to the power α . So, you can take \log here also. So, let us see, what would **what would** happen here? So, I will just do this calculation on the side in case, but this is simple calculation, but still I will just do this calculation. So, τ_0 is less than equal to 1 by **...** So, $1 - \delta n \omega$ is less than equal to $\alpha \times \log 1$ by **...** I can write also \log of $e^{-\alpha}$, but this is nothing but $1 - \delta n \omega$ by e to the power α . So, the α comes in here, and so standard \log formula.

So, this is becoming κ . I will just maintain the same **symbol** symbols. So, these are all to the **base e** \log base e ; please remember that. Nowhere in our discussion, we are using $\log 2$ base 10. Those who are thinking that is $\log 2$ base 10, please note this is $\log 2$ base e . In mathematical analysis, it is only $\log 2$ base e . (No audio from 11:55 to 12:07) this is what you have.

Now, once you have finished that calculation, we will use the standard result; \log of $1 + x$ is less than equal to x , if x is greater than equal to -1 ; this is a very standard result, once you know how to write the logarithm series, logarithm as an infinite series. So, you can take up this as the homework, and try it out. And you can find it in any standard analysis books. So, here I can take $1 + x$; x could be this $1 - \delta n \omega$.

So, what would happen is the following: \log of τ^k is less than equal to k times; x is here, $1 - \delta n \omega$; we are see this is bigger than -1 . You can choose it in such a way that this is bigger than $-1 - \delta n \omega$. So, you choose your $\delta n \omega$ in such a way that this can be made bigger than -1 ; these are positive quantities. But if I put a negative 1, I can choose in the sense that, I can choose $\delta n \omega$ in such a way that $1 - \delta n \omega$ has to be strictly less than 1; otherwise this has no meaning; $1 - \delta n \omega$ has no meaning.

So, one has to observe that this expression is meaningful, if I have this to be positive. So, **$1 - \delta n \omega$ is strictly bigger than** $1 - \delta n \omega$ is strictly bigger than 0. So, 1 is strictly bigger than $1 - \delta n \omega$ to the power ω ; so, $1 - \delta n \omega$ is strictly bigger than -1 . And so, I can apply this on this part with x is equal to $1 - \delta n \omega$, and I will get this result (no audio from 14:23 to 14:31) Now, I want τ^k ; what I actually want? I have to find the k , for which τ^k is less than equal to ϵ .

So, if τ_k is less than equal to ϵ , then \log of τ_k is less than equal to \log of ϵ . So, **we would** we would like to see, if this is less than equal to this part, if this expression, so, if k minus δn^ω plus $\alpha \log \frac{1}{\epsilon}$ is less than $\log \epsilon$. If this expression is less than equal to \log of ϵ , then it is guaranteed that $\log \tau_k$ is less than ϵ , and τ_k is also less than ϵ .

So, this will at least give me, what is the number of steps required to reach this. So, if this happens, this would imply this and that. And that these would imply τ_k to be less than ϵ . So, what I would now have is, k times minus δn^ω plus $\alpha \log \frac{1}{\epsilon}$ is less than $\log \epsilon$. Now, I will take k on the other side. You have to find this k . So, k times minus δn^ω minus $\alpha \log$ of ϵ are less than \log of ϵ . So, you see what it tells me that for any k on effort, the k for which this is less than equal to $\log \epsilon$. So, if you take any k which is bigger than that k , for which this inequality holds. Then of course for all such case, this will be satisfied

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The image shows a whiteboard with handwritten mathematical steps. At the top, there is a toolbar with various icons. The main content is as follows:

$$k \left(-\frac{\delta}{n^\omega} \right) + \alpha \log \frac{1}{\epsilon} \leq \log \epsilon$$

$$-\log \epsilon + \alpha \log \frac{1}{\epsilon} \leq k \left(\frac{\delta}{n^\omega} \right)$$

$$\Rightarrow (1 + \alpha) \log \frac{1}{\epsilon} \leq k \left(\frac{\delta}{n^\omega} \right)$$

$$\Rightarrow k \geq \frac{(1 + \alpha)}{\delta} n^\omega \log \frac{1}{\epsilon}$$

for any $k \geq \frac{(1 + \alpha)}{\delta} n^\omega \log \frac{1}{\epsilon}$, we have $\tau_k \leq \epsilon$

$$K = \frac{(1 + \alpha)}{\delta} n^\omega \log \frac{1}{\epsilon} = O \left(n^\omega \log \frac{1}{\epsilon} \right)$$

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So, this is the minimum value of k , for which this will hold, basically this expression. That is what I want to find out. So, what does it say that, if this would happen, then I would take it to this side. So, I will have k times; no it is much better, and this is not the way. It is much more simpler to do it in this way. So, I will put **log** minus $\log \epsilon$ plus $\alpha \log \frac{1}{\epsilon}$ is less than equal to k times, minus δ by small k . I am

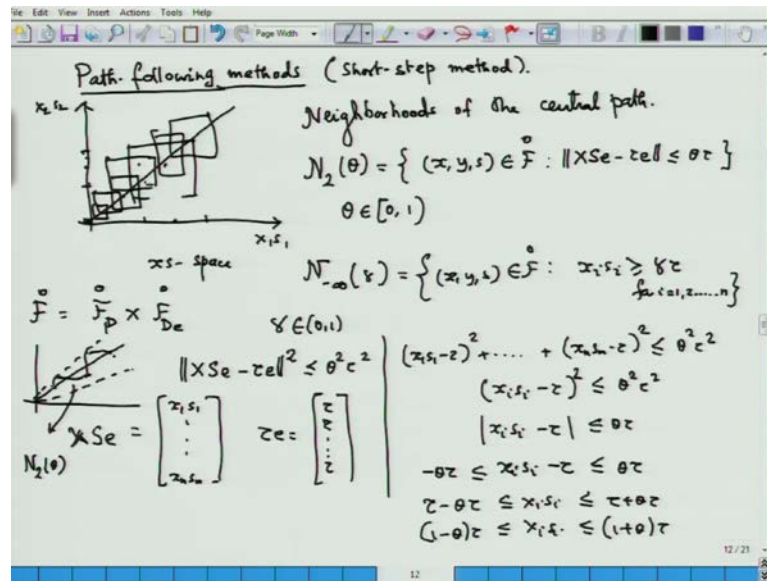
writing $\log \frac{1}{\epsilon} = n \log \frac{1}{\epsilon}$. (No audio from 17:37 to 17:48) Which implies that, I can make $\log \frac{1}{\epsilon} = n \log \frac{1}{\epsilon}$.

So, $\log \frac{1}{\epsilon} = n \log \frac{1}{\epsilon}$ is k times $\log \frac{1}{\epsilon}$. So, it implies that k is greater than equal to $\log \frac{1}{\epsilon} / \log \frac{1}{\epsilon}$. So, what you can write? Simply, you see this is now; this just tells you it is in polynomial time; you can also write $\log \frac{1}{\epsilon} = n \log \frac{1}{\epsilon}$; that does not a very big problem. Here, if you do not write in this way, here in this case, instead of this, you can just also write $\log \frac{1}{\epsilon} = n \log \frac{1}{\epsilon}$. It does not matter; it is same thing.

Now, what is this? What does it tell you? It tells you that, for any k which is bigger than this, this expression would be less than $\log \frac{1}{\epsilon}$. So, for thus τ^k would be less than equal to ϵ . So, for any k bigger than equal to $\log \frac{1}{\epsilon} / \log \frac{1}{\epsilon}$, $\tau^k < \epsilon$, we have $\tau^k < \epsilon$, because what we have shown that if this is less than this then k must be bigger than this $\log \frac{1}{\epsilon} / \log \frac{1}{\epsilon}$. So, when k is bigger than this we can just work out in the reverse version, and show that this is happening.

So, once this happens the \log of τ^k is less, than \log of ϵ . So, τ^k is less than ϵ . So, if I say k is equal to $\log \frac{1}{\epsilon} / \log \frac{1}{\epsilon}$, $\tau^k < \epsilon$; this is nothing but if I take this as the constant C , and this is nothing but order of $n \log \frac{1}{\epsilon}$. So, here I have expressed the whole thing as a function of the problem data. So, in that shows, that the problem is polynomial type. Now, can we device a algorithm can we device an algorithm which will actually follow all this property, and follow these sort of polynomial time framework. Can I device something? So, that would lead us to the study of path following methods; obviously, all the methods are path following, except the potential reduction once.

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Path following methods: So, we are first talking about the short step method. (No audio from 20:54 to 21:02) So, what does this short step method do? In the short step path following method, we have to realize one thing; we know that, if I project, say, my central path on to the x space. In this case, just we are looking at two variables, because that is the way we can visualize it; and the projection here is nothing, but the central path projection is nothing but straight line, but straight line through the origin, because $x_1 s_1$ equal to $x_2 s_2$. This is straight line which is dividing the archon into equal halves.

Now, once I solve the equation which is here, once I solve this equation, I am not exactly getting the solution of a point, which will take me to the central path or I am not exactly solving the relaxed k k t condition. So, as the result of which, what I would get is not a point on the central path, but some point near it. But we cannot allow these points to wonder of too much; that is we cannot allow this point to come near the boundary, where one of them $x_1 s_1$ can drop to 0; $x_1 s_1$, $x_2 s_2$ can drop to 0.

So, that is what we cannot allow. You cannot allow the points to come too far from the interior, and more towards the boundary. That is what the barrier does; it stops you from going to the boundary. Now, we have to then somehow force **these points** these solution points; this approximate solutions actually to stay in a region, near the central path.

So, we have to define, what is called the neighborhoods of the central path? (No audio from 22:46 to 22:58) So, that two types of neighborhoods, and one is called shorter

neighborhood; the neighborhood which allows shorter steps; there is a short step method. And another one, which allows you more flexibility, and that would be used for the long step. So, what we will use now is the following neighborhood. So, we will write down the two neighborhoods. Now this is an element of F naught, and I want to again remind you that, F naught is nothing but $F D$ cross F naught D cross F naught $D e$, which you already have seen.

So, this tells you, that the standard Euclidean norms: Finding the distances; standard distance; finding thing; distance finding; computation. So, once you know the duality measure at a particular point, find all such points which are of this form, that corresponding to this τ is the duality measure; that is $x s$, x in a product x by n ; this has to be bounded by θ times τ , and e is obviously, $1, 1, 1, 1$; the vector of ones. So, θ is a quantity, which could be 0 , but cannot become 1 . So, I am defining a neighborhood.

So, **one** another neighborhood which we will also use, which is called the infinity neighborhood or large neighborhood; what they are trying to do? They are trying to see this much; **how much you are form your duality measure**, how much you are far from the duality measure? And that is exactly what they are trying to see; how much you are far from the duality measure? To what extent you are violating this thing, this duality measure?

So, let me write down the second one. So, you take a γ which is between 0 and 1 . We will now really be concentrated on this neighborhood, rather than one I am just writing, but this is just for your information at this moment. So, when we do it, we will not mention it again. (No audio from 25:36 to 25:50) So for us, it is important to know, how this sort of neighborhoods looks like. If you look at this, how this sort of neighborhood looks like?

So, what is this? This is Euclidean norm. (No audio from 26:05 to 26:15) This square is less than θ square τ square. Now what is the meaning of this? So, what is $x s e$? Let us look at again, just to remind you this is nothing but this vector; (No audio from 26:32 to 26:42) and e is $1, 1, 1$; and τ is nothing but τ, τ, τ . τ is nothing but all components are τ . So, basically if I look at it, this would give me nothing but $x s 1$

minus tau whole square, plus $x_n s_n$ minus tau whole square, which is less than equal to theta square tau square.

Now, each of them has to be less than this quantity, because these are all non negative quantity; it is a positive quantity; it is adding up to the quantity which is less than a positive quantity. So, each of them must be less. So, what I am having is, say $\tau_i s_i$ minus $x_i s_i$ minus tau is less than theta square by theta square into tau square. So, this would just give me **that** which simply means we can just directly write down from this. This is what it means. Which means that, tau minus theta tau is less than $x_i s_i$ is less than tau plus theta tau, which means x_i into s_i for every i is $1 \pm \theta$ into tau; $1 \pm \theta$ into tau.

So, if would each i x of here $x_1 s_1$, $x_2 s_2$ is both within these two limits. So, it changes; so, what does it says? **That $x_1 s_1$** . So, $x_1 s_1$, this is say $x_1 s_1$.and this is $x_2 s_2$. So, $s_1 x_1$ is within certain limit; $x_2 s_2$ is within certain limit. So, I want all points within this; so, at particular choice of x s . So, if we take the cartesian product of these two, this is what sort **of a** for a so any point x y s , which is here has to satisfy something here. So, any point of this form $x_1 s_1$ which is say, I have given the theta, and this is my tau. **Tau is** tau also keeps on changing.

So, what you are having is that, at every point you take around that you are basically creating this sort of small squares (No audio from 29:23 to 29:30) of this sort of lengths; **the squares** the maximum length that you can have. So, it is between $1 \pm \theta$ into tau into $1 \pm \theta$ into tau. So, within that limit, everything should lie. You can keep on changing x and s s ; the tau will keep on changing. So, those lengths of the squares would change. So, basically you would have something like this. So, if you would really draw this up a bit, so, **this will be some if you** as your theta tau to become 0, the size of the squares will keep on decreasing.

So, basically if you look at if you take the edges of the square, and try to draw some sort of a tube. So, it will be something like this.

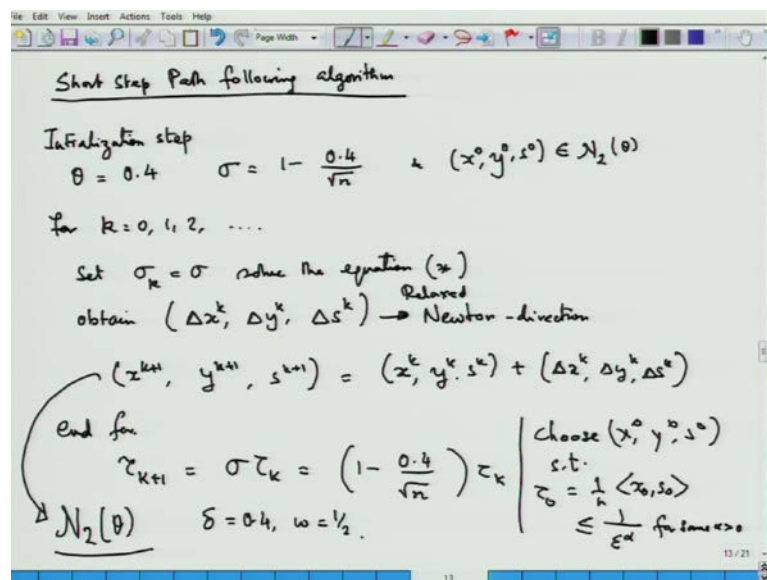
(No audio from 30:16 to 30:30)

Now this is your $N^2 \theta$. So, what is happening with this $N^2 \theta$ is that, you can **now** start from a point here; move to a point here; but **you** the algorithm should be such that.

That is what I will show you that; short step method would be such, that you are always remain with this $N_2(\theta)$, but you cannot make very large steps, **cannot take large steps**, because **you will** then there is a chance of you getting outside $N_2(\theta)$.

So, because you cannot do that, this is called the short step path following method, when you use such neighborhoods. This neighborhood would allow you to be much more flexible. So, let us just for the time being written down the short step path following algorithm. **short step path following algorithm**. So, initialization step, (No audio from 31:41 to 31:51) you will create the neighborhood $N_2(\theta)$. So, that force all the point; that is this sigma; the centering parameter is 1...

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You see **the** these choices are helpful, because with these you can choices you can show that the whole algorithm within polynomial time. And you have a starting point you have to guarantee; **not sorry** not lambda is y. This starting point has to be in these guarantee has to be taken. So, of course, you might have I cannot really find out such a thing; I can just find out a point which might be infeasible; then there is something called infeasible you have interior point methods, where you start with the initial infeasible point, but after certain success successive steps you get a feasible point.

So, just for his simple explanation **this** of this tau algorithm, we are essentially taking this fact that we are already in the feasible set. **So, now**. So, do while loop for you set. So sigma is k is not adaptive; it is not changing; it is fixed. And **solve the equation** solve the

equation star. So, this is what you have to solve the equations star. Obtain Δx_k , Δy_k , Δs_k and write. So, you are doing for loop for k equal to this to this to this. So, it is the next.

Now, here what you take is a pure Newton's step; you do not have here as scale, which is less than one; scaling parameter order control parameter with controls, how much you move from iterative x_k , y_k , s_k in the given direction. So, along the Newton direction... This is the newton direction, of often this, yes you have to relaxed Newton direction; maybe I should write relaxed newton direction. So, you add these two...

So, this is how you keep on doing, and then you have to... This is your basic algorithm, and you end for loop. So, this is you keep on doing this, till you have tau less than epsilon. Epsilon has to be put in as per as your requirement and for. So, this is the short step path following algorithm. Now, τ_{k+1} : we have shown earlier; if you remember this estimation that we did; what did we do this is the estimation. If τ_{k+1} in this case, of course, is depending on alpha; here alpha is nothing but 1. Here, if you look at this, I have no alpha here corresponding to what was here; there was an alpha here between 0 and 1; my alpha chosen it to be 1 in the short step path following method.

So, you have alpha equal to 1. So, τ_{k+1} if you again go back, and see the calculation, τ_{k+1} is $1 - (1 - \sigma)^{\tau_k}$. So, in this particular case, just let me write down. So, if when I take alpha equal to 1, this 1 and 1 will cancel, and you will have σ^{τ_k} . In this particular case τ_{k+1} is σ^{τ_k} . And what is sigma? $1 - 0.4 \sqrt{n}$ τ_k ; so, here my delta is 0.4 which is same as theta. My delta so my So, here I am having everything in the form this, and my delta is 0.4, and my omega is half. Now, what does it show? I am actually having a scenario of this result; I am actually having the basic requirement of proving the complexity or the polynomial time.

Only if I choose my x_{naught} , y_{naught} , s_{naught} in such a way, such that τ_{naught} is less than $1 - \epsilon / \alpha$, then I am done; then I already put the sequences in this form; then I am just already, I am just straight done. So, I just have to choose x_{naught} , y_{naught} , s_{naught} such that, τ_{naught} which is nothing but $1 - \epsilon / \alpha$ inner product x

naught, s naught; this τ naught is less than 1 by ϵ to the power α for some α greater than 0; ϵ is by chosen parameter, that depends on what you choose.

So, then you immediately know that you would have a polynomial time algorithm, because the basic equation is already satisfied by your choice of δ and ω . So, we have learned quite a bit of stuff today, that the general IP method; that the general format, under certain mild condition can be shown to be polynomial time. We have created an algorithm called short step path following algorithm, where all the points that we generate by solving the equation star is forced to remain in a very narrow cone, and that those points do not allow you to move very far off from the central path, but keeps you near the central path.

So, that is why it is called the short step path following algorithm, and we have showed that this short step path following algorithm can be easily made to follow the polynomial time pattern. And now, we have done our first step; polynomial time game is over. We have proved that, it is we can show it is in polynomial time. But what important stuff that we have to show, is that, when I compute x^{k+1} , y^{k+1} and s^{k+1} from x^k , y^k , s^k , then I must be sure that, this x^{k+1} , y^{k+1} and s^{k+1} is also in $N^{2\theta}$. And that is what we will discuss tomorrow, that how can we show that this is in $N^{2\theta}$. Tomorrow's discussion will concentrate on that, and then we will show some more, little bit more algorithms, before we wind up this discussion on interior point methods.

Thank you very much.