Lecture 4. Tikhonov method for equations with convex functional. Local strict convexity of the Tikhonov functional for non-linear operators.

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1 Tikhonov method for equations with convex functional

Tikhonov scheme allows to construct approximation methods for solution of equation F(x) = 0 not only for special case of operator F, when F = Ax - f.

In this lesson we will analyze more common case when functional Φ is convex, but not necessary quadratic. As before, we assume that $F = F(N_1, N_2)$. If equation have form Ax = f, then corresponding functional

$$\Phi(x) = \frac{1}{2} ||Ax - f||_{H_2}^2$$

is convex and class of operator equations which we are interested in, is not empty. This class is extension of class $F(N_1, 0)$. Differentiating Tikhonov functional

$$\Phi_{\alpha}(x) = \frac{1}{2} ||Ax - f||_{H_2}^2 + \frac{1}{2}\alpha ||x - x_0||_{H_1}^2$$

gives as

$$\Phi'_{\alpha}(x) = \Phi'(x) + \alpha(x - x_0), x \in H_1.$$

Because functional Φ is convex, we have

$$(\Phi'(x) - \Phi'(y), x - y)_{H_1} \ge 0, \forall x, y \in H_1.$$

Thus,

$$(\Phi'_{\alpha}(x) - \Phi'_{\alpha}(y), x - y)_{H_1} \ge \alpha ||x - y||^2_{H_1}, \forall x, y \in H_1.$$
(1)

Equation (2) says that functional Φ_{α} is strongly convex in H_1 for $\forall \alpha > 0$. from theory of convex optimization is known that in this case exists unique global minimizer of this problem which realizes minimum of functional Φ_{α} in H_1 .

$$\Phi_{\alpha}(x_{\alpha}) = \inf_{x \in H_1} \Phi_{\alpha}(x), \alpha > 0.$$
(2)

In optimization theory there exists a lot of minimization methods for strongly convex functionals. In all algorithms we face out with two difficulties:

- first problem is that there are not convergence estimates of x_{α} to x^* when $\alpha \to 0$.

- second problem is that iterative processes are less effective with reducing of parameter α since these processes requires increasing of optimization iterations when $\alpha \to 0$.

Let us clarify this on one example when we apply gradient method for minimizing of functional Φ_{α} :

$$x_0 \in H_1: x_{n+1} = x_n - \gamma \Phi'_{\alpha}(x_n), n = 0, 1, \dots$$
(3)

Because of (3) and necessary minimum condition

$$\Phi'_{\alpha}(x_{\alpha}) = \Phi'(x_{\alpha}) + \alpha(x_{\alpha} - x_0) = 0, n = 0, 1, \dots$$

we have

$$||x_{n+1} - x_{\alpha}||_{H_{1}}^{2} = ||x_{n} - x_{\alpha} - \gamma(\Phi_{\alpha}'(x_{n}) - \Phi'(\alpha(x_{\alpha})))||_{H_{1}}^{2}$$

= $||x_{n} - x_{\alpha}||_{H_{1}} - 2\gamma(\Phi_{\alpha}'(x_{n}) - \Phi_{\alpha}'(x_{\alpha}), x_{n} - x_{\alpha})_{H_{1}}$ (4)
+ $\gamma^{2}||\Phi_{\alpha}'(x_{n}) - \Phi_{\alpha}'(x_{\alpha})||_{H_{1}}^{2}.$

Remind that

$$\Phi'(x) = F^{*'}(x)F(x), \quad x \in H_1.$$

Following estimate takes place

$$||\Phi_{\alpha}'(x_{n}) - \Phi_{\alpha}'(x_{\alpha})||_{H_{1}}^{2} \leq \\ \leq ||\Phi'(x_{n}) - \Phi'(x_{\alpha})||_{H_{1}} + \alpha ||x_{n} - x_{\alpha}||_{H_{1}} \\ \leq ||F'^{*}(x_{n})||_{L(H_{1},H_{2})} ||F(x_{n}) - F(x_{\alpha})||_{H_{2}} \\ + ||F'^{*}(x_{n}) - F'^{*}(x_{\alpha})||_{L(H_{2},H_{1})} ||F(x_{\alpha})||_{H_{2}} \\ + \alpha ||x_{n} - x_{\alpha}||_{H_{1}} \\ \leq (N_{1}^{2} + N_{2}||F(x_{\alpha})||_{H_{2}} + \alpha) ||x_{n} - x_{\alpha}||_{H_{1}}.$$

$$(5)$$

By definition of x_{α}

$$\Phi_{\alpha}(x_{\alpha}) = \frac{1}{2} ||F(x_{\alpha})||_{H_{2}}^{2} + \frac{\alpha}{2} ||x_{\alpha} - x_{0}||_{H_{1}}^{2} \le \Phi_{\alpha}(x_{x_{o}}^{*} - x_{0})_{H_{1}}^{2}$$

and thus

$$||F(x_{\alpha})||_{H_{2}} \le \sqrt{\alpha} \Phi_{\alpha} (x_{x_{o}}^{*} - x_{0})_{H_{1}}$$
(6)

From estimates (5) and (6) follows that

$$\begin{aligned} |\Phi_{\alpha}'(x_n) - \Phi_{\alpha}'(x_{\alpha})||_{H_1}^2 &\leq L ||x_n - x_{\alpha}||_{H_1}, \\ L &= N_1^2 + N_2 ||x_{x_0}^* - x_0||_{H_1} \sqrt{\alpha} + \alpha. \end{aligned}$$
(7)

Substituting estimates (5) and (7) in (4) and using (2) we can get

$$||x_{n+1} - x_{\alpha}||_{H_1}^2 \le \sqrt{1 - 2\gamma\alpha + \gamma^2 L^2} ||x_n - x_{\alpha}||_{H_1}.$$
(8)

We see that for $\forall \gamma > 0$ in (8) $1 - 2\gamma\alpha + \gamma^2 L^2 > 0$. Value of γ we should choose like that expression $1 - 2\gamma\alpha + \gamma^2 L^2$ should be small. We can require that

$$1 - 2\gamma\alpha + \gamma^2 L^2 < 1.$$

For this it is sufficient that following condition should be fullfilled

$$0 < \gamma < \frac{2\alpha}{L^2}.$$
(9)

In the case that (9) is true we have

$$||x_n - x_\alpha||_{H_1}^2 \le \sqrt{1 - 2\gamma\alpha + \gamma^2 L^2} ||x_0 - x_\alpha||_{H_1} q^n(\alpha),$$

$$q(\alpha) = \sqrt{1 - 2\gamma\alpha + \gamma^2 L^2}, \quad q(\alpha) \in (0, 1).$$
(10)

From (10) follows that defined by (3) sequence x_n converges to x_α as geometric progression. Then minimal value its achieved at $\gamma = \alpha L^{-2}$ and corresponding minimal value of $1 - 2\gamma\alpha + \gamma^2 L^2$ is equal to $1 - \alpha^2/L^2$. Thus, when $\alpha \to 0$ then $q(\alpha) \to 1$ for any choice of γ from condition (9). Because of this we need to make more iterations in minimizing of our functional.

1.1 Convexity of Tikhonov functional for linear operators

Let H_1 and H_2 be two real valued Hilbert spaces. Denote norms in these spaces as $\|\bullet\|_1$ and $\|\bullet\|_2$. Let the operator $F: H_1 \to H_2$. Consider a general equation

$$F(x) = 0, x \in H_1.$$
 (11)

An important case of this equation is

$$F(x) = Ax - f, A \in \mathbb{L}(H_1, H_2); x \in H_1, f \in H_2.$$
(12)

Thus, we will discuss now the linear operator equation

$$Ax = f, x \in H_1, A \in \mathbb{L}(H_1, H_2), f \in H_2.$$
(13)

In other words, $A: H_1 \to H_2$ is a bounded linear operator. In the case of ill-posed problems A is a compact operator. Therefore, a bounded inverse A^{-1} does not exist. Thus, we need to consider the Tikhonov functional Φ_{α} to approximately solve equation (3), where

$$\Phi_{\alpha}(x) = \frac{1}{2} \|Ax - f\|_{2}^{2} + \frac{\alpha}{2} \|x - x_{0}\|_{1}^{2}.$$
 (A)

Here x_0 is a certain a priori chosen vector. In computational practice usually x_0 is a good approximation for the correct solution.

We now establish the strict convexity of $\Phi_{\alpha}(x)$. This is an important property. Indeed, it is well known that a strictly convex functional can have at most one point of minimum. Now, if dim $H_1 < \infty$, then this minimum exists. Indeed, it is clear that

$$\lim_{\left\|x\right\|_{1}\to\infty}\Phi_{\alpha}\left(x\right)=\infty$$

On the other hand,

$$\Phi_{\alpha}(0) = \frac{\|f\|_{2}^{2}}{2} + \frac{\alpha \|x_{0}\|_{1}^{2}}{2}.$$

In other words, $\Phi_{\alpha}(0)$ has a certain bounded value. Hence, by the Weierstrass theorem a minimizer x_{α} of $\Phi_{\alpha}(x)$ exists and by Theorem 1 this minimizer is unique: the uniqueness is because $\Phi_{\alpha}(x)$ is strictly convex.

Definition 1. Let $U \subseteq H_1$ be convex set and the operator $G: U \to H_2$ (G is not necessary linear). The operator G has the Frechet derivative $G'(x) \in L(H_1, H_2)$ at the point $x \in U$ if the following representation is valid

$$G(x+h) - G(x) = G'(x)(h) + o_x(||h||_1), \forall h \in H_1,$$
(100)

where

$$\lim_{\|h\|_{1} \to 0} \frac{\|o_{x}(\|h\|_{1})\|_{2}}{\|h\|_{1}} = 0.$$
(101)

In other words, the difference G(x+h) - G(x) is a linear operator G'(x)(h) plus a function, whose norm us much smaller than $||h||_1$ for sufficiently small $||h||_1$. This is the full similarity with the conventional definition of the differential.

Furthermore, it follows from (100) and (101) that in order to figure out the Frechet derivative G'(x)(h) at the point x, one should consider the difference G(x+h) - G(x) and consider the linear, with respect to h, part of this difference. So, this part is a good candidate for being the Frechet derivative: do not forget that we also need to prove (101).

Definition. Let $U \subseteq H_1$ be convex set and let the functional $\Phi : U \to \mathbb{R}$ has the Frechet derivative $\Phi'(x)$ for all $x \in U$. This functional is called strictly convex on the set U with the strict convexity parameter κ if

$$\Phi(x+y) - \Phi(x) - \Phi'(x)(y) \ge \kappa \|y\|_1^2, \forall x, y \in U \text{ such that } x+y \in U.$$
(B)

Theorem 1. Suppose that the operator A in (13) is linear and bounded. Then the functional Φ_{α} is strictly convex on the space H_1 with the strict convexity parameter $\alpha/2$.

Proof. First, we consider the Frechet derivative $\Phi'_{\alpha}(x)$ of the functional $\Phi_{\alpha}(x)$. So, by the above rule, we need to consider the difference $\Phi_{\alpha}(x+h) - \Phi_{\alpha}(x)$, single out its linear part (with respect to h) and prove (101) for the rest. Let $(,)_1$ and $(,)_2$ be scalar products in H_1 and H_2 respectively. Since we have real valued Hilbert spaces, then $(a,b)_k = (b,a)_k, \forall a, b \in H_k, k = 1, 2$. Hence,

$$\Phi_{\alpha}(x+h) - \Phi_{\alpha}(x) = \frac{1}{2} (Ax + Ah, Ax + Ah)_{2} + \frac{\alpha}{2} (x+h, x+h)_{1}$$

$$-\frac{1}{2}(Ax, Ax)_{2} - \frac{\alpha}{2}(x, x)_{1}$$
(C)
= $[(Ax, Ah)_{2} + \alpha (x, h)_{1}] + \left[\frac{1}{2}(Ah, Ah)_{2} + \frac{\alpha}{2}(h, h)_{1}\right].$

Since A is bounded operator, than

$$\left|\frac{1}{2}(Ah,Ah)_{2} + \frac{\alpha}{2}(h,h)_{1}\right| \leq \frac{1}{2}(\|A\| + \alpha) \|h\|_{1}^{2}.$$

Hence,

$$\begin{split} &\lim_{\|h\|_{1}\to 0} \frac{1}{\|h\|_{1}} \left\{ \left[\frac{1}{2} \left(Ah, Ah\right)_{2} + \frac{\alpha}{2} \left(h, h\right)_{1} \right] \right\} \\ &\leq \quad \frac{1}{2} \lim_{\|h\|_{1}\to 0} \left(\|A\| + \alpha\right) \|h\|_{1} = 0. \end{split}$$

Therefore, it follows from (100), (101) and (C) that

$$\begin{array}{rcl} \Phi_{\alpha}^{\prime}\left(x\right)\left(h\right) &=& \left(Ax,Ah\right)_{2}+\alpha\left(x,h\right)_{1}\\ \text{or } \Phi_{\alpha}^{\prime}\left(x\right)\left(h\right) &=& \left(\left(AA^{*}+\alpha\right)x,h\right)_{1}. \end{array}$$

Hence, $\Phi'_{\alpha}(x) = (AA^* + \alpha I) x$, where I is the identity operator on H_1 .

Now, we are ready to use (B) to prove the strict convexity of Φ_{α} . We have

$$\Phi_{\alpha} (x + y) - \Phi_{\alpha} (x) - \Phi_{\alpha}' (x) (y)$$

= $\frac{1}{2} (Ax + Ay, Ax + Ay)_2 + \frac{\alpha}{2} (x + y, x + y)_1$
 $-\frac{1}{2} (Ax, Ax)_2 - \frac{\alpha}{2} (x, x)_1 - (Ax, Ay)_2 - \alpha (x, y)_1$
= $\frac{1}{2} (Ay, Ay)_2 + \frac{\alpha}{2} (y, y)_1 \ge \frac{\alpha}{2} ||y||_1^2.$

We have used here the fact that $(Ay, Ay)_2 \ge 0$. \Box

2 Existence of minimizer

As before we assume that there exists Frechet derivative of functional $\Phi_{\alpha}(x)$ and $F \in (N_1, N_2)$.

We assume that

$$||F(x^*)|| \le \delta, 0 < \delta < 1.$$

$$\tag{14}$$

We also assume that the operator F has the Fechet derivative F'(x) for $x \in V_1(x^*) = \{||x - x^*|| < 1\}$ and this derivative is Lipschitz continuous or

$$||F'(x)|| \le N_1, ||F'(x) - F'(y)|| < N_2||x - y||.$$
(15)

We now assume that

$$||x_{glob} - x^*|| \leq \delta^{\mu_1}, \mu_1 = const. \in (0, 1),$$
 (16)

$$\alpha = \delta^{\mu_2}, \mu_2 = const. \in (0, \min(\mu_1, 2(1-\mu_1)))$$
(17)

Lemma A minimizer x_{α} of the functional $\Phi_{\alpha}(x)$ on the space H_1 exists for any value of the regularization parameter α . For any r > 0 denote $V_r(x_{\alpha}) = \{x \in H_1 : ||x - x_{\alpha}|| < r\}$. Assume that conditions (16), (17) hold. Then $x_{glob} \in V_{\sqrt{2}\delta^{\mu_1}}(x_{\alpha})$ and $x^* \in V_{(1+\sqrt{2})\delta^{\mu_1}}(x_{\alpha})$. Let $\beta_1 \in (0, 1)$ be any number. Then there exists a sufficiently small number $\delta_0 = \delta_0(\mu_1, \mu_2, \beta_1) \in (0, 1)$ such that if $\delta \in (0, \delta_0)$, then $x^*, x_{glob} \in V_{\beta_1 \alpha}(x_{\alpha})$.

Proof.

Since dim $H < \infty$, then $\lim_{\|x\|\to\infty} \Phi_{\alpha}(x) = \infty$ implies the existence of a minimizer x_{α} . Since $\Phi_{\alpha}(x_{\alpha}) \leq \Phi_{\alpha}(x^{*})$ then

$$\Phi_{\alpha}(x^{*}) = \Phi(x^{*}) + \frac{\alpha}{2} ||x^{*} - x_{glob}|| = \frac{1}{2} ||F(x^{*})||^{2} + \frac{\alpha}{2} ||x^{*} - x_{glob}|| \le \frac{\delta^{2} + \alpha \delta^{2\mu_{1}}}{2}, \quad (18)$$

and then

$$\Phi_{\alpha}(x_{\alpha}) \le \Phi_{\alpha}(x^{*}) = \frac{\delta^{2} + \alpha \delta^{2\mu_{1}}}{2}$$
(19)

and thus

$$\Phi_{\alpha}\left(x_{\alpha}\right) \leq \frac{\delta^{2} + \alpha \delta^{2\mu_{1}}}{2} < \alpha \delta^{2\mu_{1}} \tag{20}$$

Why
$$\frac{\delta^2 + \alpha \delta^{2\mu_1}}{2} < \alpha \delta^{2\mu_1}$$
?

Since $\alpha = \delta^{\mu_2}$, then

$$\frac{\alpha\delta^{2\mu_1}}{\delta^2} = \delta^{\mu_2 - 2(1-\mu_1)}$$

By (17) $\mu_2 < 2(1 - \mu_1)$. Hence $\mu_2 - 2(1 - \mu_1) < 0$. Hence,

$$\lim_{\delta \to 0} \delta^{\mu_2 - 2(1 - \mu_1)} = \infty. \tag{D}$$

But in Lemma we talk about sufficiently small δ_0 . Hence, it follows from (D) that if δ_0 is sufficiently small, then for $\delta \in (0, \delta_0)$ we have

$$\frac{\alpha\delta^{2\mu_1}}{\delta^2} = \delta^{\mu_2 - 2(1-\mu_1)} > 1 \to \alpha\delta^{2\mu_1} > \delta^2.$$

Then, by the definition of $\Phi_{\alpha}(x_{\alpha})$ we have

$$\Phi_{\alpha}(x_{\alpha}) = \Phi(x_{\alpha}) + \frac{\alpha}{2} ||x_{\alpha} - x_{glob}||^{2} = \frac{1}{2} ||F(x_{\alpha})||^{2} + \frac{\alpha}{2} ||x_{\alpha} - x_{glob}||^{2} < \alpha \delta^{2\mu_{1}}.$$
 (21)

It follows from this equation that

$$\frac{1}{2}||F(x_{\alpha})||^{2} < \alpha \delta^{2\mu_{1}}.$$
(22)

and

$$\frac{\alpha}{2}||x_{\alpha} - x_{glob}||^2 < \alpha \delta^{2\mu_1}.$$
(23)

From the last inequality we get that

$$||x_{\alpha} - x_{glob}|| < \sqrt{2}\delta^{\mu_1}.$$
(24)

Using this estimate, we obtain

$$||x_{\alpha} - x_{glob}|| \leq ||x_{\alpha} - x_{glob}|| + ||x_{0} - x^{*}|| < \sqrt{2}\delta^{\mu_{1}} + \delta^{\mu_{1}} = (1 + \sqrt{2})\delta^{\mu_{1}} < \beta_{1}\alpha = \beta_{1}\delta^{\mu_{2}}.$$
(25)

Why $(1 + \sqrt{2})\delta^{\mu_1} < \beta_1 \alpha = \beta_1 \delta^{\mu_2}$? Since $\mu_2 < \mu_1$, then

$$\lim_{\delta \to 0} \frac{\beta_1 \delta^{\mu_2}}{(1+\sqrt{2})\delta^{\mu_1}} = \infty$$

and δ is sufficiently small. Thus

$$\frac{\beta_1\delta^{\mu_2}}{(1+\sqrt{2})\delta^{\mu_1}}>1$$

and thus

$$\beta_1 \delta^{\mu_2} > (1 + \sqrt{2}) \delta^{\mu_1}.$$

Theorem Assume that conditions (16), (17) hold. Then there exists numbers $\beta_1 = \beta_1 (N_1, N_2) \in (0, 1)$ and $\delta_1 = \delta_1 (\mu_1, \mu_2, N_2, \beta_1) \in (0, 1)$ depending only on listed parameters such that if $\rho = \beta_1 \alpha$, then for any $\delta \in (0, \delta_1)$ the functional J_{α} is strictly convex in the neighborhood $V_{\rho}(x_{\alpha})$ of the point x_{α} with the strict convexity parameter $\kappa = \alpha/4$. Furthermore, by Lemma 2.1 points $x_{glob}, x^* \in V_{\rho}(x_{\alpha})$.

Proof. Let $\beta_1 \in (0, 1)$ be the number which we will choose below in this proof, $\rho = \beta_1 \alpha$ and $x, y \in V_{\rho}(x_{\alpha})$ be two arbitrary points. By (2.9)

$$(J'_{\alpha}(x) - J'_{\alpha}(y), x - y) = \alpha ||x - y||^{2} + (F'^{*}(x) F(x) - F'^{*}(y) F(y), x - y)$$

$$= \alpha ||x - y||^{2} + (F'^{*}(x) F(x) - F'^{*}(x) F(y), x - y) \quad (26)$$

$$+ (F'^{*}(x) F(y) - F'^{*}(y) F(y), x - y).$$

Denote $A_1 = (F'^*(x) F(x) - F'^*(x) F(y), x - y), A_2 = (F'^*(x) F(y) - F'^*(y) F(y), x - y)$ and estimate A_1, A_2 from the below. Since $A_1 = A_1 - (F'^*(x) F'(x) (x - y), x - y) + (F'^*(x) F'(x) (x - y), x - y)$, then we have

$$A_{1} = (F'^{*}(x) F(x) - F'^{*}(x) F(y), x - y) - (F'^{*}(x) F'(x) (x - y), x - y) + (F'^{*}(x) F'(x) (x - y), x - y) = (F'^{*}(x) (F(x) - F(y), x - y) - (F'^{*}(x) F'(x) (x - y), x - y) + (F'^{*}(x) F'(x) (x - y), x - y),$$

then using main theorem of calculus

$$F(x) - F(y) = \int_{0}^{1} (F'(y + \theta(x - y)))(x - y) d\theta$$

we can rewrite A_1 as

$$A_{1} = (F'^{*}(x) \int_{0}^{1} (F'(y + \theta(x - y)) - F'(x))(x - y) d\theta, x - y) + (F'^{*}(x) F'(x)(x - y), x - y).$$

We obtain

$$\begin{aligned} |F'^*(x)\int_0^1 [F'(y+\theta(x-y))-F'(x)](x-y)d\theta, x-y| \\ &\leq \|F'(x)\|\int_0^1 \|[F'(y+\theta(x-y))-F'(x)](x-y)\|d\theta \cdot \|x-y\| \\ &\leq N_1N_2\int_0^1 (y+\theta(x-y)-x)(x-y)d\theta \cdot \|x-y\| \\ &\leq \frac{1}{2}N_1N_2 \|x-y\|^3 \,. \end{aligned}$$
$$z &= \int_0^1 [F'(y+\theta(x-y))-F'(x)](x-y)d\theta. \end{aligned}$$

Hence,

$$\left| \left(F'^{*}(x) \int_{0}^{1} [F'(y + \theta(x - y)) - F'(x)](x - y)d\theta, x - y \right) \right|$$

$$= |(F'^*(x)z, x - y)| \le ||F'^*(x)|| \cdot ||z|| \cdot ||x - y||$$

= ||F'(x)|| \cdot ||z|| \cdot ||x - y||.

We have used here that for every bounded linear operator A we have $||A^*|| = ||A||$. Also,

$$(F'^*(x) F'(x) (x - y), x - y) = (F'(x) (x - y), F'(x) (x - y))_2 = ||F'(x) (x - y)||_2^2 \ge 0.$$

Hence, $A_1 \ge N_1 N_2 ||x - y||^2 / 2$. Now we estimate A_2 ,

$$|A_{2}| \leq ||F(y)||_{2} ||F'(x) - F'(y)|| ||x - y|| \leq N_{2} ||x - y||^{2} ||F(y)||_{2}$$

Since

$$\|F(y)\|_{2} \leq \|F(y) - F(x_{\alpha})\|_{2} + \|F(x_{\alpha})\|_{2} \leq N_{1} \|y - x_{\alpha}\| + \|F(x_{\alpha})\|_{2}, \qquad (27)$$

since

$$||F(y) - F(x_{\alpha})||_{2} \le (\max_{x \in V_{1}(x^{*})} ||F'(x)||)||y - x_{\alpha}|| \le N_{1}||y - x_{\alpha}||,$$
(28)

then

$$A_{2}| \leq N_{2} \|x - y\|^{2} \left(N_{1} \|y - x_{\alpha}\| + \|F(x_{\alpha})\|_{2}\right).$$

$$(29)$$

By (14), (15) and Lemma $\|F(x_{\alpha})\|_{2} \leq \|F(x_{\alpha}) - F(x^{*})\|_{2} + \delta \leq \alpha \beta_{1} N_{1} + \delta$. Hence,

$$A_{2} \geq -N_{2} \|x - y\|^{2} (N_{1} \|y - x_{\alpha}\| + \alpha \beta_{1} N_{1} + \delta).$$
(30)

Combining this with (26) and the above estimate for A_1 , we obtain

$$\left(\Phi_{\alpha}'\left(x\right) - \Phi_{\alpha}'\left(y\right), x - y\right) \ge \tag{31}$$

$$\|x-y\|^{2} \left[\alpha - \frac{N_{1}N_{2}}{2} \|x-y\| - N_{1}N_{2} \|y-x_{\alpha}\| - N_{2} \left(N_{1}\alpha\beta_{1} + \delta\right) \right].$$

We have by Lemma

$$\frac{||x-y||}{2} = \frac{||x-x_{\alpha}+x_{\alpha}-y||}{2} \le \frac{\alpha\beta_{1}+\alpha\beta_{1}}{2} = \alpha\beta_{1}.$$
(32)

and thus

$$N_{1}N_{2}\frac{\|x-y\|}{2} + N_{1}N_{2}\|y-x_{\alpha}\| + N_{2}\left(N_{1}\alpha\beta_{1}+\delta\right)$$

$$\leq \alpha\beta_{1}N_{2}N_{1} + \alpha\beta_{1}N_{2}N_{1} + \alpha\beta_{1}N_{2}N_{1} + N_{2}\delta$$

$$\leq 3\alpha\beta_{1}N_{2}N_{1} + N_{2}\delta.$$
(33)

Choose $\beta_1 = \beta_1 (N_1, N_2) \in (0, 1)$ such that $3\beta_1 N_2 N_1 \leq 1/4$. Given this β_1 , choose $\delta_1 = \delta_1 (\mu_1, \mu_2, N_1, N_2) \in (0, 1)$ so small that

$$N_2\delta < \delta^{\mu_2}/4 = \alpha/4 \tag{34}$$

and $2\delta^{\mu_1} < \beta_1 \delta^{\mu_2} = \beta_1 \alpha, \forall \delta \in (0, \delta_1)$. Then we have that $\alpha/4 + \alpha/4 = \alpha/2$ in (33) and thus from (31) we have $\alpha - \alpha/2 = \alpha/2$.