

REDUCTION FACTOR OF MULTIGRID ITERATIONS FOR ELLIPTIC PROBLEMS

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1. Introduction

Multigrid method has been used widely to solve elliptic problems because of its applicability to many class of problems and fast convergence ([1],[3], [9], [10], [11], [12]). The estimate of convergence rate of multigrid is one of the main objectives of the multigrid analysis([1], [2], [5], [6], [7], [8]). In many problems, the convergence rate depends on the regularity of the solution([5], [6], [8]). In this paper, we present an improved estimate of reduction factor of multigrid iteration based on the proof in[6].

2. Elliptic problems in \mathbf{R}^2

Let Ω be a polygonal domain in \mathbf{R}^2 and let

$$(2.1) \quad \begin{aligned} -Lu &= f \text{ in } \Omega \\ u &= 0 \text{ on } \partial\Omega, \end{aligned}$$

where L is a uniformly elliptic operator and $f \in L^2(\Omega)$. We further assume the solution u satisfies the elliptic regularity: $u \in H_0^{1+\alpha}(\Omega)$. Let $S_h(\Omega)$ be a finite dimensional subspace of $H_0^1(\Omega)$, say, the space of continuous, piecewise linear functions on some triangulation of Ω . We use standard finite element method to solve (2.1), i.e.,

$$(2.2) \quad A(u, \phi) = (f, \phi), \quad \forall \phi \in S_h(\Omega).$$

Received June 28, 1993.

1991 AMS subject classification: Primary 65N30; secondary 65F10.

Key words: Multigrid method, elliptic problems, reduction factor.

The author was partially supported by the Applied Mathematics Research Center at Korea Advanced Institute of Science and Technology.

Let $M_1 \subset \cdots \subset M_J = S_h(\Omega)$ be a nested sequence of subspaces of $S_h(\Omega)$. Let $(\cdot, \cdot)_k$ be the discrete L^2 inner product on M_k . To define multigrid algorithm we need some operators. For $k = 1, \dots, J$ define $A_k : M_k \rightarrow M_k, P_k : M_J \rightarrow M_k, P_{k-1}^0 : M_k \rightarrow M_{k-1}$ via

$$\begin{aligned} (A_k u, \phi)_k &= A(u, \phi), \quad \forall \phi \in M_k, \\ (A_{k-1} P_{k-1} u, \phi)_{k-1} &= A(u, \phi), \quad \forall \phi \in M_{k-1}, \\ (P_{k-1}^0 u, \phi)_{k-1} &= (u, \phi)_k, \quad \forall \phi \in M_{k-1}. \end{aligned}$$

and define the multigrid algorithm as usual:

Multigrid Algorithm Set $S_1 = A_1^{-1}$. Assume S_{k-1} has been defined and define $S_k(g)$ as follows:

- (1) Set $x^0 = 0$ and $c^0 = 0$.
- (2) Define x^l for $l = 1, \dots, m$ by

$$(2.3) \quad x^l = x^{l-1} + R_k(g - A_k x^{l-1}).$$

where R_k is any symmetric smoother.

- (3) Define $x^{m+1} = x^m + c^p$ where $c^i, i = 1, \dots, p$ is given as

$$(2.4) \quad c^i = c^{i-1} + S_{k-1}[P_{k-1}^0(g - A_k x^m) - A_{k-1} c^{i-1}].$$

- (4) Define x^l for $l = m + 2, \dots, 2m + 1$ by

$$(2.5) \quad x^l = x^{l-1} + R_k(g - A_k x^{l-1}).$$

- (5) Set $S_k(g) = x^{2m+1}$.

One can also define an algorithm without (2.5). Set $N_1 = A_1^{-1}$. Assume N_{k-1} has been defined and define $N_k(g)$ as follows:

Nonsymmetric Algorithm

- (1) Set $x^0 = 0$ and $c^0 = 0$.
- (2) Define x^l for $l = 1, \dots, m$ by

$$(2.6) \quad x^l = x^{l-1} + R_k(g - A_k x^{l-1}).$$

- (3) Define c^p for, $i = 1, \dots, p$ given as

$$(2.7) \quad c^i = c^{i-1} + N_{k-1}[P_{k-1}^0(g - A_k x^m) - A_{k-1} c^{i-1}].$$

- (4) Set $N_k(g) = x^m + c^p$.

Let c satisfy $P_{k-1}^0(g - A_k x^m) - A_{k-1}c = 0$. Then

$$\begin{aligned} c^i - c &= c^{i-1} - c - N_{k-1}A_{k-1}(c^{i-1} - c) \\ &= (I - N_{k-1}A_{k-1})(c^{i-1} - c) \\ c^p - c &= (I - N_{k-1}A_{k-1})^p(c^{i-1} - c) \end{aligned}$$

therefore

$$c^p = [(I - N_{k-1}A_{k-1})^p]A_{k-1}^{-1}P_{k-1}^0A_k(x - x^m).$$

Also

$$\begin{aligned} x - x^m &= x - x^{m-1} - R_kA_k(x - x^{m-1}) \\ &= (I - R_kA_k)(x - x^{m-1}) \\ &= K_k^m x. \end{aligned}$$

Thus, from $P_{k-1}^0A_k = A_{k-1}P_{k-1}$, and $Ax = g$,

$$\begin{aligned} (I - N_kA_k)x &= x - N_k g = x - x^m - c^p \\ &= K_k^m x - [I - (I - N_{k-1}A_{k-1})^p]A_{k-1}^{-1}P_{k-1}^0A_k(x - x^m) \\ &= (I - [I - (I - N_{k-1}A_{k-1})^p]P_{k-1})P_{k-1}K_k^m x \\ &= [I - P_{k-1} + (I - N_{k-1}A_{k-1})^p P_{k-1}]K_k^m x. \end{aligned}$$

For the symmetric case, we have similarly,

$$(2.8) \quad I - S_kA_k = K_k^m [(I - P_{k-1}) + (I - S_{k-1}A_{k-1})^p P_{k-1}]K_k^m.$$

Thus

$$(2.9) \quad \begin{aligned} A((I - S_kA_k)u, v) &= A((I - P_{k-1})K_k^m u, K_k^m v) \\ &+ (A(I - S_{k-1}A_{k-1})^p P_{k-1}K_k^m u, K_k^m v). \end{aligned}$$

We also find the relation between symmetric multigrid algorithm and nonsymmetric multigrid algorithm:

$$A((I - S_kA_k)u, u) = A((I - N_kA_k)u, (I - N_kA_k)u).$$

3. Estimates of convergence factor

We shall show that for symmetric algorithm

$$(3.1) \quad A((I - S_k A_k)u, u) \leq \delta_k A(u, u)$$

and

$$(3.2) \quad A((I - N_k A_k)u, (I - N_k A_k)u) \leq \delta_k A(u, u)$$

for nonsymmetric algorithm. For the proof we need two assumptions.

First of all, the following regularity and approximation property:

$$(3.3) \quad A((I - P_{k-1})u, u) \leq C_\alpha^2 \left(\frac{\|A_k u\|_k^2}{\lambda_k} \right)^\alpha A(u, u)^{1-\alpha}, \quad u \in M_k,$$

where λ_k is the largest eigenvalue of A_k . This follows from the regularity of the solution of the underlying differential equation and the approximation property of the subspaces M_k .

Next, we have from the smoothing property of R_k ,

$$(3.4) \quad \frac{\|u\|_k^2}{\lambda_k} \leq C_R(R_k u, u)_k, \quad u \in M_k.$$

Now the following result is from [6].

THEOREM A. *Assume (3.3) and (3.4). Then $S_k(\cdot)$ defined with $p = 1$ satisfies (3.1) with*

$$\delta_k = 1 - \varepsilon_k$$

$$\varepsilon_k = \frac{m^\alpha}{m^\alpha + M_\alpha(j+k)^{(1-\alpha)/\alpha}}$$

where $\tilde{M}_\alpha = \left(\frac{1+j}{j}\right)^s \frac{C_R(\alpha C_\alpha^2)^{1/\alpha}}{2}$ and

$$s = \begin{cases} \frac{1-\alpha}{\alpha}, & \alpha \geq \frac{1}{2} \\ \left(\frac{1-\alpha}{\alpha}\right)^2, & \alpha < \frac{1}{2}. \end{cases}$$

Set

$$M_\alpha = \left(\frac{1 + \tilde{M}_\alpha}{\tilde{M}_\alpha} \right)^{\frac{1-\alpha}{\alpha}}.$$

In this proof, however, the convergence factor δ_k depends heavily on α and $\delta_k \rightarrow 1$ very fast as $\alpha \rightarrow 0$. In this paper, we try to improve above result by a more careful analysis. We have under same assumptions,

THEOREM 1. $S_J(\cdot)$ defined with $p = 1$ satisfies (3.1) with

$$\delta_k = 1 - \varepsilon_k$$

$$(3.5) \quad \varepsilon_k = \frac{m^\alpha}{m^\alpha + M_\alpha(j+k)^{(1-\alpha)}}$$

where

$$(3.6) \quad M_\alpha = S_\alpha^{1-\alpha} (C_\alpha^2) \left(\alpha \frac{C_R}{2} \right)^\alpha$$

and S_α, D_α are quantities satisfying

$$(3.7) \quad S_\alpha = \frac{1}{D_\alpha} \left(\frac{A-1}{A} \right)^{-\alpha}.$$

In this result we have several choices for S_α . We shall see some of the examples later.

Proof. From (2.9) and the induction hypothesis,

$$(3.8) \quad \begin{aligned} & A((I - P_{k-1})u, u) \\ & \leq A((I - P_{k-1})K_k^m u, K_k^m u) + \delta_{k-1} A(P_{k-1}K_k^m u, K_k^m u). \\ & = (1 - \delta_{k-1})A((I - P_{k-1})K_k^m u, K_k^m u) + \delta_{k-1} A(P_{k-1}K_k^m u, K_k^m u). \end{aligned}$$

As in the proof of Theorem A in [6],

$$\begin{aligned} & \leq [(1 - \delta_{k-1})C_\alpha^2(1 - \alpha)\gamma_k^{-\alpha/(1-\alpha)} + \delta_{k-1}]A(K_k^{2m}u, u) \\ & \quad + (1 - \delta_{k-1})C_\alpha^2 C_R \frac{\alpha}{2m} \gamma_k A((I - K_k^{2m})u, u). \end{aligned}$$

Then (3.1) will follow if we choose γ_k so that

$$(3.9) \quad (1 - \delta_{k-1})C_\alpha^2(1 - \alpha)\gamma_k^{-\alpha/(1-\alpha)} + \delta_{k-1} \leq \delta_k$$

$$(3.10) \quad \text{and } (1 - \delta_{k-1})C_\alpha^2 C_R \frac{\alpha}{2m} \gamma_k \leq \delta_k.$$

Set γ_k so that

$$(3.11) \quad (1 - \delta_{k-1})C_\alpha^2 C_R \frac{\alpha}{2m} \gamma_k = \delta_{k-1}.$$

Since $\delta_k \geq \delta_{k-1}$, (3.10) follows as soon as (3.11) holds. We need to check (3.9) which is equivalent to

$$(3.12) \quad (1 - \delta_{k-1})C_\alpha^2(1 - \alpha)\gamma_k^{-\alpha/(1-\alpha)} \leq \delta_k - \delta_{k-1}.$$

Let $D(k) = m^\alpha + M_\alpha(j+k)^{1-\alpha}$ and let

$$(3.13) \quad 1 - \delta_k = \varepsilon_k = \frac{m^\alpha}{m^\alpha + M_\alpha(j+k)^{1-\alpha}} = \frac{m^\alpha}{D(k)}, \quad \alpha < 1.$$

Then

$$(3.14) \quad \delta_k - \delta_{k-1} = \varepsilon_{k-1} - \varepsilon_k = \frac{M_\alpha m^\alpha}{D(k)D(k-1)} [(j+k)^{1-\alpha} - (j+k-1)^{1-\alpha}].$$

With $A = j+k$, we see that

$$(3.15) \quad [A^{1-\alpha} - (A-1)^{1-\alpha}] = A^{-\alpha} \left[1 - \left(1 - \frac{1}{A}\right)^{1-\alpha} \right] \geq (1-\alpha)A^{-\alpha}.$$

Since

$$(3.16) \quad \varepsilon_{k-1} - \varepsilon_k = \frac{(1-\alpha)M_\alpha m^\alpha}{D(k)D(k-1)} (j+k)^{-\alpha},$$

The left side of (3.12) is

$$(3.17) \quad (1 - \delta_{k-1})^{\frac{1}{1-\alpha}} (C_\alpha^2)^{\frac{1}{1-\alpha}} (1 - \alpha) \left(\frac{\alpha C_R}{2m} \right)^{\frac{\alpha}{1-\alpha}} \cdot \left[\frac{D(k-1)}{M_\alpha(j+k-1)^{1-\alpha}} \right]^{\frac{\alpha}{1-\alpha}}$$

We want to show (3.17) \leq (3.16) which means

$$(3.18) \quad \frac{1}{D(k-1)} (C_\alpha^2)^{\frac{1}{1-\alpha}} \left(\frac{\alpha C_R}{2M_\alpha} \right)^{\frac{\alpha}{1-\alpha}} (j+k-1)^{-\alpha} \leq \frac{M_\alpha m^\alpha (j+k)^{-\alpha}}{D(k)D(k-1)}$$

or

$$(3.19) \quad (C_\alpha^2)^{\frac{1}{1-\alpha}} \left(\frac{\alpha C_R}{2M_\alpha} \right)^{\frac{\alpha}{1-\alpha}} (j+k-1)^{-\alpha} \leq \frac{M_\alpha m^\alpha (j+k)^{-\alpha}}{D(k)}$$

Let $\widetilde{M}_\alpha^{1-\alpha} = (C_\alpha^2)^{\frac{1}{1-\alpha}} \left(\frac{\alpha C_R}{2}\right)^{\frac{\alpha}{1-\alpha}}$.

Then it suffices to show

$$(3.20) \quad \begin{aligned} \widetilde{M}_\alpha^{1-\alpha} &\leq M_\alpha^{\frac{1}{1-\alpha}} m^\alpha \left(\frac{A-1}{A}\right)^\alpha / D(k), \\ \widetilde{M}_\alpha^{1-\alpha} [m^\alpha + M_\alpha(j+k-1)^{1-\alpha}] &\leq M_\alpha^{\frac{1}{1-\alpha}} m^\alpha \left(\frac{A-1}{A}\right)^\alpha \end{aligned}$$

Let $\widetilde{M}_\alpha^{1-\alpha} = M_\alpha^{\frac{1}{1-\alpha}} \left(\frac{A-1}{A}\right)^\alpha D_\alpha$. Then

$$D_\alpha [m^\alpha + M_\alpha(j+k-1)^{1-\alpha}] \leq m^\alpha$$

And hence

$$(3.21) \quad D_\alpha M_\alpha(j+k-1)^{1-\alpha} \leq m^\alpha - D_\alpha m^\alpha = (1 - D_\alpha) m^\alpha$$

Set

$$M_\alpha^{\frac{1}{1-\alpha}} = S_\alpha (C_\alpha^2)^{\frac{1}{1-\alpha}} \left(\frac{\alpha C_R}{2}\right)^{\frac{\alpha}{1-\alpha}}.$$

Then from

$$\widetilde{M}_\alpha^{1-\alpha} = M_\alpha^{\frac{1}{1-\alpha}} \left(\frac{A-1}{A}\right)^\alpha D_\alpha$$

we have $D_\alpha = \frac{1}{S_\alpha} \left(\frac{A-1}{A}\right)^{-\alpha}$. Hence (3.21) is equivalent to

$$(3.22) \quad \frac{1}{S_\alpha} \left(\frac{A-1}{A}\right)^{-\alpha} S_\alpha^{1-\alpha} C_\alpha^2 \left(\frac{\alpha C_R}{2}\right)^\alpha (A-1)^{1-\alpha} \leq (1 - D_\alpha) m^\alpha$$

It is equivalent to

$$(3.23) \quad S_\alpha^{-\alpha} \left(\frac{A-1}{A}\right)^{-\alpha} C_\alpha^2 \left(\frac{\alpha C_R}{2}\right)^\alpha (A-1)^{1-\alpha} \leq (1 - D_\alpha) m^\alpha$$

which holds if S_α is sufficiently large.

Now we give examples of S_α and D_α for which (3.23) holds. Let S_α be any number such that $S_\alpha^\alpha \rightarrow S_0$ as $\alpha \rightarrow 0$, where S_0 is some large number greater than zero. Then $D_\alpha \rightarrow S_0^{-1}$ and since $C_\alpha \rightarrow C_0 = 1$, (3.23) becomes

$$\frac{(A-1)}{S_0} \leq (1 - S_0^{-1}).$$

Which holds if $S_0 \geq A$.

EXAMPLE 1. Let $S_\alpha^\alpha = (1 + 2\alpha \ln A)^{1/\alpha}$. Then $S_\alpha^\alpha \rightarrow A^2$ and (3.23) holds for α sufficiently small.

EXAMPLE 2. Let

$$S_\alpha^\alpha = \left(\frac{A-1}{A}\right)^{-\alpha} C_\alpha^2 C \left(\frac{\alpha C_R}{2}\right)^\alpha (A-1)^{1-\alpha}$$

for some large constant C .

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