

Convergence of Jacobi, GS and SOR(ω). (1)

When do these methods converge? How fast?

Definition. A is strictly row diagonally dominant if $|a_{ii}| > \sum_{j \neq i} |a_{ij}|$.

Theorem. If A is strictly row diagonally dominant, then both Jacobi and GS converge. GS converges "faster" in the sense that $\|R_{GS}\|_\infty \leq \|R_J\|_\infty < 1$.

Proof. We have $R_J = L + U = (D^{-1}(L + \tilde{U}))$ and $R_{GS} = (I - L)^{-1}U$. We want to show

$$\|R_{GS}\| = \|(R_{GS})\vec{1}\|_\infty \leq \|(R_J)\vec{1}\|_\infty = \|R_J\|_\infty.$$

where $\vec{1}$ is the vector of all ones.

(Thus $A\vec{1}$ is the vector of row sums, etc.)

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In fact, we show the stronger componentwise inequality (for all i)

$$(\|R_{Gj}\|_1)_i \leq (\|R_j\|_1)_i$$

Now we have

$$\|(\mathbf{I} - L)^{-1} u\|_1 \leq \|(\mathbf{I} - L)^{-1}\| \|u\|_1$$

triangle inequality,

since these are sums of absolute values and lhs is abs. vals. of sums.

strictly

Now L is \uparrow lower triangular, and so

$$\begin{bmatrix} * & * & & \\ * & 0 & & \\ 0 & * & * & * \\ 0 & * & * & * \end{bmatrix} \quad \begin{bmatrix} * & * & * & * \\ 0 & * & * & * \\ 0 & 0 & * & * \\ 0 & 0 & 0 & * \end{bmatrix}$$

$$\begin{bmatrix} 0 & x & \cdots & x \\ 0 & 0 & \ddots & x \\ \vdots & \vdots & \ddots & x \\ 0 & 0 & \cdots & 0 \end{bmatrix} \begin{bmatrix} 0 & x & \cdots & x \\ 0 & 0 & \ddots & x \\ 0 & 0 & 0 & \ddots & x \\ 0 & 0 & 0 & 0 & \ddots & x \\ \vdots & \vdots & \ddots & \vdots & \ddots & 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 & x & x \\ 0 & 0 & 0 & x \\ 0 & 0 & 0 & 0 & \ddots & x \\ 0 & 0 & 0 & 0 & 0 & 0 \\ \vdots & \vdots & \ddots & \vdots & \ddots & 0 \end{bmatrix}$$

we can compute that the subdiagonals empty out as we take powers of L .

In fact, $L^n = 0$. So ~~$(\mathbf{I} - L)^{-1}$~~ Now we can expand

$$(\mathbf{I} - L)^{-1} = \sum_{i=0}^{\infty} L^i$$

as usual and truncate the series at L^{n-1} .

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We learn that

$$(I - L)^{-1} = \sum_{i=0}^{n-1} L^i$$

so our previous

$$\|(I - L)^{-1}\|_1 \leq \left\| \sum_{i=0}^{n-1} L^i \right\|_1$$

$$\leq \sum_{i=0}^{n-1} \|L\|^i \|u\|_1$$

again, triangle inequality.

$$= (1 - \|L\|)^{-1} \cdot \|u\|_1$$

since $\|L\|^n = 0$ as well.

So this means it is ok to prove (componentwise)

$$(I - \|L\|)^{-1} \|u\|_1 \leq (\|L\| + \|u\|)_1$$

Now $(I - \|L\|)^{-1}$ is a matrix with nonnegative entries, ~~as~~ since it's $\sum \|L\|^i$. So if we can prove

$$\|u\|_1 \leq (I - \|L\|)(\|L\| + \|u\|)_1$$

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we could obtain the inequality by multiplying both sides by $(I - |L|)^{-1}$.

So we want to show

$$|U| \vec{1} \leq (I - |L|)(|L| + |U|) \vec{1}$$

or

$$\begin{aligned} 0 &\leq [(I - |L|)(|L| + |U|) \vec{1}] \vec{1} \\ &\leq [|L| + |U| - |L|^2 - |L||U|] \vec{1} \\ &\leq [|L| - |L|^2 - |L||U|] \vec{1} \\ &\leq |L| (I - |L| - |U|) \vec{1} \end{aligned}$$

But again, this will be true if

$$0 \leq (I - |L| - |U|) \vec{1}$$

OK

Now $I \cdot \vec{1} = \vec{1}$, while $(|L| + |U|) \cdot \vec{1}$ = the vector of row sums of $|L| + |U|$. But such a sum

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is equal to $\sum_{\substack{j=1 \\ j \neq k}}^n |a_{kj}|$, which is < 1

by the assumption that A is strictly row diagonally dominant.

Now in fact $|L| + |U| = R_j$ so if A is S.R.D.D. then $\|R_j\|_\infty = \max \text{row sum}$ of $R_j < 1$. Thus we have (unwinding this whole mess) that

$$\|R_{GS}\|_\infty \leq \|R_j\|_\infty$$

as claimed. Now we have shown that in this operator norm

$$\|R_{GS}\|_\infty \leq \|R_j\|_\infty < 1,$$

so GS and Jacobi both converge. \square

In most cases, we can settle for

Definition. A is weakly diagonally

dominant if $|a_{jj}| \geq \sum_{j \neq k} |a_{jk}|$ and strict
inequality occurs at least once.

We need a technical hypothesis: think

about the variables x_i as nodes in a
graph. x_i and x_j are connecting connected

if $a_{ij} \neq 0$, this allows information to flow
from x_i to x_j as we iterate. ~~If the~~

If the graph is disconnected, it could
split into a bunch of independent problems,
some of which might not have any
strict inequalities.

Theorem. (see Demmel, Thm 6.3) If A is
irreducible (i.e. the graph is connected) and
weakly diagonally dominant, GS and Jacobi
converge and $\rho(R_{GS}) < \rho(R_f) < 1$.

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What about $SOR(\omega)$? Recall

$$R_{SOR(\omega)} = (I - \omega L)^{-1}((1-\omega)I + \omega U).$$

Theorem. $\rho(R_{SOR(\omega)}) \geq |\omega - 1|$. Thus
 $0 < \omega < 2$ for convergence.

Proof. Consider the characteristic polynomial of $R_{SOR(\omega)}$:

$$\begin{aligned}\Phi(\lambda) &= \det(\lambda I - R_{SOR(\omega)}) \\ &= \det(\lambda I - (I - \omega L)^{-1}((1-\omega)I + \omega U)) \\ &= \det(\underbrace{(I - \omega L)}_{\text{has det 1}} (\lambda I - (I - \omega L)^{-1}((1-\omega)I + \omega U))) \\ &= \det((\lambda + \omega - 1)I - \omega \lambda L - \omega U)\end{aligned}$$

So the constant term in $\Phi(\lambda)$, is given by

$$\begin{aligned}\Phi(0) &= \det((\omega - 1)I - \omega U) \\ &= \det((\omega - 1)I) = (\omega - 1)^n\end{aligned}$$

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Now the constant term in any polynomial is the product of the roots, so this means that if $\lambda_1, \dots, \lambda_n$ are the eigenvalues of $R_{SOR}(\omega)$, then

$$\prod_{i=1}^n \lambda_i = (\omega - 1)^n$$

So one of the λ_i must ~~be~~ have $|\lambda_i| \geq |\omega - 1|$, as desired. \square

It turns out to be the case that when A is symmetric positive definite,

Theorem. A s.p.d. $\Rightarrow \rho(R_{SOR}(\omega)) < 1$
 for all $\omega \in (0, 2)$ so $SOR(\omega)$ converges.

(For $\omega = 1$, note this shows GS also converges.)

This depends on an amazing theorem.

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Spectral Mapping Theorem.

Suppose p is a rational function.

If A is a square matrix with eigenvalues $\{\lambda_i\}$ then $P(A)$ has eigenvalues $\{P(\lambda_i)\}$.

Proof (of SOR(ω) theorem) The splitting for SOR(ω) is

$$\begin{aligned} A &= M - K \\ &= \frac{1}{\omega}(D - \omega \tilde{L}) - \frac{1}{\omega}((1-\omega)D + \omega \tilde{U}) \end{aligned}$$

So let

$$Q = A^{-1}(2M - A).$$

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We claim $\operatorname{Re} \lambda_i(Q) > 0$ for all eigenvalues λ_i of Q . Suppose $Qx = \lambda x$.

Then

$$(2M - A)x = \lambda Ax$$

or (as a scalar equation)

$$x^*(2M - A)x = \lambda x^*Ax$$

Now the conjugate transpose of this equation is

$$(x^*(2M - A)x)^* = \lambda^* (x^*Ax)^*$$

or

$$x^*(2M^* - A)x = \lambda^* (x^*Ax)$$

Since A is symmetric (and real). So if we add this to the equation above,

$$x^*(2M + 2M^* - 2A)x = (\lambda^* + \lambda)(x^*Ax)$$

or

$$x^*(M + M^* - A)x = (\operatorname{Re} \lambda)(x^*Ax).$$

We can now solve for

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$$\operatorname{Re} \lambda = \frac{x^*(M+M^*-A)x}{x^*Ax}$$

Now

$$\begin{aligned} M+M^* &= \omega^{-1}(D-\omega\tilde{L}) + \omega^{-1}(D-\omega\tilde{L}^*) \\ &= \omega^{-1}(2D - \omega\tilde{L} - \omega\tilde{U}) \\ &\quad \leftarrow \text{b/c } A \text{ symmetric} \\ &= \omega^{-1}((2-\omega)D + \omega D - \omega\tilde{L} - \omega\tilde{U}) \\ &= \omega^{-1}((2-\omega)D + \omega A) \\ &= \cancel{\omega} \frac{(2-\omega)}{\omega} D + A \end{aligned}$$

so

$$M+M^*-A = \left(\frac{2}{\omega}-1\right)D$$

We know $\omega < 2$, so this is a positive multiple of D .

Thus $(M^* + M^* - A)$ is positive definite, ⑫
as is A , so

$$\operatorname{Re} \lambda = \frac{x^*(M+M^*-A)x}{x^*Ax} > 0.$$

Now we claim $(Q-I)(Q+I)^{-1} = R_{SOR(\omega)}$.

This can just be checked:

\leftarrow do so \rightarrow

So we have for each eigenvalue of Q ,
the corresponding eigenvalue of R has
~~absolute value~~ complex norm

$$\left| \frac{\lambda-1}{\lambda+1} \right| = \left| \frac{(\operatorname{Re} \lambda - 1)^2 + (\operatorname{Im} \lambda)^2}{(\operatorname{Re} \lambda + 1)^2 + (\operatorname{Im} \lambda)^2} \right|^{\frac{1}{2}}$$

which is < 1 because $\operatorname{Re} \lambda > 0$. This proves that $R_{SOR(\omega)}$ has spectral radius less than 1 as desired.