

1. Two iterations of Jacobi's method gives the following results.

(a) $\mathbf{x}^{(2)} = (1.2500000, -1.3333333, 0.2000000)^t$

(b) $\mathbf{x}^{(2)} = (-1.0000000, 1.0000000, -1.3333333)^t$

(c) $\mathbf{x}^{(2)} = (-0.5208333, -0.04166667, -0.2166667, 0.4166667)^t$

(d) $\mathbf{x}^{(2)} = (0.6875, 1.125, 0.6875, 1.375, 0.5625, 1.375)^t$

2. Two iterations of Jacobi's method gives the following results.

(a) $\mathbf{x}^{(2)} = (0.1428571, -0.3571429, 0.4285714)^t$

(b) $\mathbf{x}^{(2)} = (0.97, 0.91, 0.74)^t$

(c) $\mathbf{x}^{(2)} = (-0.65, 1.65, -0.4, -2.475)^t$

(d) $\mathbf{x}^{(2)} = (1.325, -1.6, 1.6, 1.675, 2.425)^t$

3. Two iterations of the Gauss-Seidel method give the following results.

(a) $\mathbf{x}^{(2)} = (0.1111111, -0.2222222, 0.6190476)^t$

(b) $\mathbf{x}^{(2)} = (0.979, 0.9495, 0.7899)^t$

(c) $\mathbf{x}^{(2)} = (-0.5, 2.64, -0.336875, -2.267375)^t$

(d) $\mathbf{x}^{(2)} = (1.189063, -1.521354, 1.862396, 1.882526, 2.255645)^t$

4. Two iterations of the Gauss-Seidel method give the following results.

(a) $\mathbf{x}^{(2)} = (1.250000000, -0.9166666667, 0.06666666666)^t$

(b) $\mathbf{x}^{(2)} = (-1.666666667, 1.333333334, -0.8888888894)^t$

(c) $\mathbf{x}^{(2)} = (-0.625, 0, -0.225, 0.6166667)^t$

(d) $\mathbf{x}^{(2)} = (0.6875, 1.546875, 0.7929688, 1.71875, 0.7226563, 1.878906)^t$

5. Jacobi's Algorithm gives the following results.

(a) $\mathbf{x}^{(10)} = (1.447642384, -0.8355647882, -0.0450226618)^t$

(b) $\mathbf{x}^{(21)} = (-1.45485795, 1.45485795, -0.72704396)^t$

(c) $\mathbf{x}^{(12)} = (-0.75205599, 0.04027028, -0.28025957, 0.69008536)^t$

(d) $\mathbf{x}^{(9)} = (0.35705566, 1.42852883, 0.35705566, 1.57141113, 0.28552246, 1.57141113)^t$

6. Jacobi's Algorithm gives the following results.

(a) $\mathbf{x}^{(9)} = (0.03510079, -0.23663751, 0.65812732)^t$

(b) $\mathbf{x}^{(6)} = (0.9957250, 0.9577750, 0.7914500)^t$

(c) $\mathbf{x}^{(21)} = (-0.79710581, 2.79517067, -0.25939578, -2.25179299)^t$

(d) $\mathbf{x}^{(12)} = (0.7870883, -1.003036, 1.866048, 1.912449, 1.985707)^t$

7. The Gauss-Seidel Algorithm gives the following results.

- (a) $\mathbf{x}^{(6)} = (1.447816350, -0.8358173037, -0.0447996186)^t$
- (b) $\mathbf{x}^{(8)} = (-1.45480420, 1.45441316, -0.72720658)^t$
- (c) $\mathbf{x}^{(8)} = (-0.7531763, 0.04101049, -0.2807047, 0.6916305)^t$
- (d) $\mathbf{x}^{(6)} = (0.35713196, 1.42856598, 0.35714149, 1.57140350, 0.28570175, 1.57142544)^t$

8. The Gauss-Seidel Algorithm gives the following results.

- (a) $\mathbf{x}^{(6)} = (0.03535107, -0.2367886, 0.6577590)^t$
- (b) $\mathbf{x}^{(4)} = (0.9957475, 0.9578738, 0.7915748)^t$
- (c) $\mathbf{x}^{(9)} = (-0.79691476, 2.79461827, -0.25918081, -2.25183616)^t$
- (d) $\mathbf{x}^{(7)} = (0.7866825, -1.002719, 1.866283, 1.912562, 1.989790)^t$

9. (a)

$$T_j = \begin{bmatrix} 0 & \frac{1}{2} & -\frac{1}{2} \\ -1 & 0 & -1 \\ \frac{1}{2} & \frac{1}{2} & 0 \end{bmatrix} \quad \text{and} \quad \det(\lambda I - T_j) = \lambda^3 + \frac{5}{4}\lambda.$$

Thus, the eigenvalues of T_j are 0 and $\pm \frac{\sqrt{5}}{2}i$, so $\rho(T_j) = \frac{\sqrt{5}}{2} > 1$.

- (b) $\mathbf{x}^{(25)} = (-20.827873, 2.0000000, -22.827873)^t$

(c)

$$T_g = \begin{bmatrix} 0 & \frac{1}{2} & -\frac{1}{2} \\ 0 & -\frac{1}{2} & -\frac{1}{2} \\ 0 & 0 & -\frac{1}{2} \end{bmatrix} \quad \text{and} \quad \det(\lambda I - T_g) = \lambda \left(\lambda + \frac{1}{2} \right)^2.$$

Thus, the eigenvalues of T_g are 0, $-\frac{1}{2}$, and $-\frac{1}{2}$; and $\rho(T_g) = \frac{1}{2}$.

- (d) $\mathbf{x}^{(23)} = (1.0000023, 1.9999975, -1.0000001)^t$ is within 10^{-5} in the l_∞ norm.

10. (a) $T_j = \begin{bmatrix} 0 & -2 & 2 \\ -1 & 0 & -1 \\ -2 & -2 & 0 \end{bmatrix}$ and $\det(\lambda I - T_j) = \lambda^3$, so $\rho(T_j) = 0$.

- (b) $\mathbf{x}^{(4)} = (1.00000000, 2.00000000, -1.00000000)^t$ is within 10^{-5} in the l_∞ norm.

(c) $T_g = \begin{bmatrix} 0 & -2 & 2 \\ 0 & 2 & -3 \\ 0 & 0 & 2 \end{bmatrix}$ and $\det(\lambda I - T_g) = \lambda(\lambda - 2)^2$, so $\rho(T_g) = 2$.

- (d) $\mathbf{x}^{(25)} = (1.30 \times 10^9, -1.325 \times 10^9, 3.355 \times 10^7)^t$

11. (a) A is not strictly diagonally dominant.
 (b)

$$T_g = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 0 & 0.75 \\ 0 & 0 & -0.625 \end{bmatrix} \quad \text{and} \quad \rho(T_g) = 0.625.$$

Since T_g is convergent, the Gauss-Seidel method will converge.

- (c) With $\mathbf{x}^{(0)} = (0, 0, 0)^t$, $\mathbf{x}^{(13)} = (0.89751310, -0.80186518, 0.70155431)^t$
 (d) $\rho(T_g) = 1.375$. Since T_g is not convergent, the Gauss-Seidel method will not converge.
12. (a) A is not strictly diagonally dominant.

(b) We have

$$T_j = \begin{bmatrix} 0 & 0 & 1 \\ 0.5 & 0 & 0.25 \\ -1 & 0.5 & 0 \end{bmatrix} \quad \text{and} \quad \rho(T_j) = 0.97210521.$$

Since T_j is convergent, the Jacobi method will converge.

- (c) With $\mathbf{x}^{(0)} = (0, 0, 0)^t$, $\mathbf{x}^{(187)} = (0.90222655, -0.79595242, 0.69281316)^t$
 (d) $\rho(T_j) = 1.39331779371$. Since T_j is not convergent, the Jacobi method will not converge.
13. (a) Subtract $\mathbf{x} = T\mathbf{x} + \mathbf{c}$ from $\mathbf{x}^{(k)} = T\mathbf{x}^{(k-1)} + \mathbf{c}$ to obtain $\mathbf{x}^{(k)} - \mathbf{x} = T(\mathbf{x}^{(k-1)} - \mathbf{x})$. Thus,

$$\|\mathbf{x}^{(k)} - \mathbf{x}\| \leq \|T\| \|\mathbf{x}^{(k-1)} - \mathbf{x}\|.$$

Inductively, we have

$$\|\mathbf{x}^{(k)} - \mathbf{x}\| \leq \|T\|^k \|\mathbf{x}^{(0)} - \mathbf{x}\|.$$

The remainder of the proof is similar to the proof of Corollary 2.5.

- (b) The last column has no entry when $\|T\|_\infty = 1$.

	$\ \mathbf{x}^{(2)} - \mathbf{x}\ _\infty$	$\ T\ _\infty$	$\ T\ _\infty^2 \ \mathbf{x}^{(0)} - \mathbf{x}\ _\infty$	$\frac{\ T\ _\infty^2}{1 - \ T\ _\infty} \ \mathbf{x}^{(1)} - \mathbf{x}^{(0)}\ _\infty$
1 (a)	0.22932	0.857143	0.48335	2.9388
1 (b)	0.051579	0.3	0.089621	0.11571
1 (c)	1.1453	0.9	2.2642	20.25
1 (d)	0.27511	1	0.75342	
1 (e)	0.59743	1	1.9897	
1 (f)	0.875	0.75	1.125	3.375

14. The matrix $T_j = (t_{ik})$ has entries given by

$$t_{ik} = \begin{cases} 0, & i = k \text{ for } 1 \leq i \leq n \text{ and } 1 \leq k \leq n \\ -\frac{a_{ik}}{a_{ii}}, & i \neq k \text{ for } 1 \leq i \leq n \text{ and } 1 \leq k \leq n. \end{cases}$$

Since A is strictly diagonally dominant, $\|T_j\|_\infty = \max_{1 \leq i \leq n} \sum_{\substack{k=1 \\ k \neq i}}^n \left| \frac{a_{ik}}{a_{ii}} \right| < 1$.

15.

	Jacobi 33 iterations	Gauss-Seidel 8 iterations		Jacobi 33 iterations	Gauss-Seidel 8 iterations
x_1	1.53873501	1.53873270	x_{41}	0.02185033	0.02184781
x_2	0.73142167	0.73141966	x_{42}	0.02133203	0.02132965
x_3	0.10797136	0.10796931	x_{43}	0.02083782	0.02083545
x_4	0.17328530	0.17328340	x_{44}	0.02036585	0.02036360
x_5	0.04055865	0.04055595	x_{45}	0.01991483	0.01991261
x_6	0.08525019	0.08524787	x_{46}	0.01948325	0.01948113
x_7	0.16645040	0.16644711	x_{47}	0.01907002	0.01906793
x_8	0.12198156	0.12197878	x_{48}	0.01867387	0.01867187
x_9	0.10125265	0.10124911	x_{49}	0.01829386	0.01829190
x_{10}	0.09045966	0.09045662	x_{50}	0.71792896	0.01792707
x_{11}	0.07203172	0.07202785	x_{51}	0.01757833	0.01757648
x_{12}	0.07026597	0.07026266	x_{52}	0.01724113	0.01723933
x_{13}	0.06875835	0.06875421	x_{53}	0.01691660	0.01691487
x_{14}	0.06324659	0.06324307	x_{54}	0.01660406	0.01660237
x_{15}	0.05971510	0.05971083	x_{55}	0.01630279	0.01630127
x_{16}	0.05571199	0.05570834	x_{56}	0.01601230	0.01601082
x_{17}	0.05187851	0.05187416	x_{57}	0.01573198	0.01573087
x_{18}	0.04924911	0.04924537	x_{58}	0.01546129	0.01546020
x_{19}	0.04678213	0.04677776	x_{59}	0.01519990	0.01519909
x_{20}	0.04448679	0.04448303	x_{60}	0.01494704	0.01494626
x_{21}	0.04246924	0.04246493	x_{61}	0.01470181	0.01470085
x_{22}	0.04053818	0.04053444	x_{62}	0.01446510	0.01446417
x_{23}	0.03877273	0.03876852	x_{63}	0.01423556	0.01423437
x_{24}	0.03718190	0.03717822	x_{64}	0.01401350	0.01401233
x_{25}	0.03570858	0.03570451	x_{65}	0.01380328	0.01380234
x_{26}	0.03435107	0.03434748	x_{66}	0.01359448	0.01359356
x_{27}	0.03309542	0.03309152	x_{67}	0.01338495	0.01338434
x_{28}	0.03192212	0.03191866	x_{68}	0.01318840	0.01318780
x_{29}	0.03083007	0.03082637	x_{69}	0.01297174	0.01297109
x_{30}	0.02980997	0.02980666	x_{70}	0.01278663	0.01278598
x_{31}	0.02885510	0.02885160	x_{71}	0.01270328	0.01270263
x_{32}	0.02795937	0.02795621	x_{72}	0.01252719	0.01252656
x_{33}	0.02711787	0.02711458	x_{73}	0.01237700	0.01237656
x_{34}	0.02632478	0.02632179	x_{74}	0.01221009	0.01220965
x_{35}	0.02557705	0.02557397	x_{75}	0.01129043	0.01129009
x_{36}	0.02487017	0.02486733	x_{76}	0.01114138	0.01114104
x_{37}	0.02420147	0.02419858	x_{77}	0.01217337	0.01217312
x_{38}	0.02356750	0.02356482	x_{78}	0.01201771	0.01201746
x_{39}	0.02296603	0.02296333	x_{79}	0.01542910	0.01542896
x_{40}	0.02239424	0.02239171	x_{80}	0.01523810	0.01523796

16. (a) We have $P_0 = 1$, so the equation $P_1 = \frac{1}{2}P_0 + \frac{1}{2}P_2$ gives $P_1 - \frac{1}{2}P_2 = \frac{1}{2}$. Since $P_i = \frac{1}{2}P_{i-1} + \frac{1}{2}P_{i+1}$, we have $-\frac{1}{2}P_{i-1} + P_i - \frac{1}{2}P_{i+1} = 0$, for $i = 2, \dots, n-2$. Finally, since $P_n = 0$ and $P_{n-1} = \frac{1}{2}P_{n-2} + \frac{1}{2}P_n$, we have $-\frac{1}{2}P_{n-2} + P_{n-1} = 0$. This gives the linear system.

(b) The solution vector is $(0.89996431, 0.79993544, 0.69991549, 0.59990552, 0.49990552, 0.39991454, 0.29993086, 0.19995223, 0.09997611)^t$, using 86 iterations with a tolerance 1.00×10^{-5} in l_∞ with the Gauss-Seidel method.

The solution vector is $(0.96289774, 0.92595527, 0.88925042, 0.85285897, 0.81685427, 0.78130672, 0.74628346, 0.71184798, 0.67805979, 0.64497421, 0.61264206, 0.58110953, 0.55041801, 0.52060401, 0.49169906, 0.46372973, 0.43671763, 0.41067944, 0.38562707, 0.36156768, 0.33850391, 0.31643400, 0.29535198, 0.27524791, 0.25610805, 0.23791514, 0.22064859, 0.20428475, 0.18879715, 0.17415669, 0.16033195, 0.14728936, 0.13499341, 0.12340690, 0.11249111, 0.10220596, 0.09251023, 0.08336165, 0.07471709, 0.06653267, 0.05876386, 0.05136562, 0.04429243, 0.03749843, 0.03093747, 0.02456315, 0.01832893, 0.01218814, 0.00609407)^t$, using 231 iterations with tolerance 1.00×10^{-3} in l_∞ with the Gauss-Seidel method.

The solution vector is $(0.96305854, 0.92627494, 0.88972613, 0.85348706, 0.81763026, 0.78222543, 0.74733909, 0.71303418, 0.67936983, 0.64640101, 0.61417841, 0.58274816, 0.55215178, 0.52242602, 0.49360287, 0.46570950, 0.43876832, 0.41279701, 0.38780868, 0.36381196, 0.34081114, 0.31880642, 0.29779408, 0.27776668, 0.25871338, 0.24062014, 0.22346997, 0.20724328, 0.19191807, 0.17747025, 0.16387393, 0.15110162, 0.13912457, 0.12791297, 0.11743622, 0.10766312, 0.09856216, 0.09010163, 0.08224988, 0.07497547, 0.06824731, 0.06203481, 0.05630801, 0.05103770, 0.04619548, 0.04175387, 0.03768638, 0.03396754, 0.03057293, 0.02747926, 0.02466435, 0.02210715, 0.01978772, 0.01768725, 0.01578806, 0.01407350, 0.01252803, 0.01113710, 0.00988718, 0.00876568, 0.00776092, 0.00686210, 0.00605926, 0.00534321, 0.00470552, 0.00413844, 0.00363490, 0.00318842, 0.00279312, 0.00244363, 0.00213509, 0.00186308, 0.00162362, 0.00141311, 0.00122831, 0.00106630, 0.00092447, 0.00080047, 0.00069221, 0.00059781, 0.00051560, 0.00044409, 0.00038197, 0.00032806, 0.00028132, 0.00024082, 0.00020575, 0.00017539, 0.00014909, 0.00012629, 0.00010648, 0.00008920, 0.00007405, 0.00006067, 0.00004871, 0.00003787, 0.00002786, 0.00001839, 0.00000919)^t$, using 233 iterations with tolerance 1.00×10^{-3} in l_∞ norm with the Gauss-Seidel method.

- (c) The equations are $P_i = \alpha P_{i-1} + (1 - \alpha)P_{i+1}$, for $i = 1, 2, \dots, n - 1$, and the linear system becomes

$$\begin{bmatrix} 1 & \alpha - 1 & 0 & \cdots & 0 \\ -\alpha & 1 & \alpha - 1 & \ddots & \vdots \\ 0 & -\alpha & \ddots & \ddots & 0 \\ \vdots & \ddots & \ddots & \ddots & \alpha - 1 \\ 0 & \cdots & 0 & -\alpha & 1 \end{bmatrix} \begin{bmatrix} P_1 \\ P_2 \\ \vdots \\ \vdots \\ P_{n-1} \end{bmatrix} = \begin{bmatrix} \alpha \\ 0 \\ \vdots \\ \vdots \\ 0 \end{bmatrix}$$

- (d) The solution vector is $(0.49947985, 0.24922901, 0.12411164, 0.06155895, 0.03028662, 0.01465286, 0.00683728, 0.00293009, 0.00097670)^t$,

using 35 iterations with tolerance 1.00×10^{-5} in l_∞ norm with the Gauss-Seidel method.

The solution vector is $(4.9995328 \times 10^{-1}, 2.4993967 \times 10^{-1}, 1.2494215 \times 10^{-1}, 6.2451172 \times 10^{-2}, 3.1211719 \times 10^{-2}, 1.5596448 \times 10^{-2}, 7.7919757 \times 10^{-3}, 3.8919193 \times 10^{-3}, 1.9433556 \times 10^{-3}, 9.7003673 \times 10^{-4}, 4.8399876 \times 10^{-4}, 2.4137415 \times 10^{-4}, 1.2030832 \times 10^{-4}, 5.9927321 \times 10^{-5}, 2.9829260 \times 10^{-5}, 1.4835803 \times 10^{-5}, 7.3721140 \times 10^{-6}, 3.6597037 \times 10^{-6}, 1.8148167 \times 10^{-6}, 8.9890229 \times 10^{-7}, 4.4467752 \times 10^{-7}, 2.1968048 \times 10^{-7}, 1.0837093 \times 10^{-7}, 5.3379165 \times 10^{-8}, 2.6250177 \times 10^{-8}, 1.2887232 \times 10^{-8}, 6.3156953 \times 10^{-9}, 3.0894829 \times 10^{-9}, 1.5084277 \times 10^{-9}, 7.3503876 \times 10^{-10}, 3.5745232 \times 10^{-10}, 1.7347035 \times 10^{-10}, 8.4006105 \times 10^{-11}, 4.0593470 \times 10^{-11}, 1.9572418 \times 10^{-11}, 9.4158798 \times 10^{-12}, 4.5195197 \times 10^{-12}, 2.1643465 \times 10^{-12}, 1.0340770 \times 10^{-12}, 4.9289757 \times 10^{-13}, 2.3437677 \times 10^{-13}, 1.1116644 \times 10^{-13}, 5.2577488 \times 10^{-14}, 2.4776509 \times 10^{-14}, 1.1608190 \times 10^{-14}, 5.3767458 \times 10^{-15}, 2.4249977 \times 10^{-15}, 1.0192489 \times 10^{-15}, 3.3974965 \times 10^{-16})^t$,

using 40 iterations with tolerance 1.00×10^{-5} in l_∞ norm with the Gauss-Seidel method.

The solution vector is $(4.9995328 \times 10^{-1}, 2.4993967 \times 10^{-1}, 1.2494215 \times 10^{-1}, 6.2451172 \times 10^{-2}, 3.1211719 \times 10^{-2}, 1.5596448 \times 10^{-2}, 7.7919757 \times 10^{-3}, 3.8919193 \times 10^{-3}, 1.9433556 \times 10^{-3}, 9.7003673 \times 10^{-4}, 4.8399876 \times 10^{-4}, 2.4137415 \times 10^{-4}, 1.2030832 \times 10^{-4}, 5.9927321 \times 10^{-5}, 2.9829260 \times 10^{-5}, 1.4835803 \times 10^{-5}, 7.3721140 \times 10^{-6}, 3.6597037 \times 10^{-6}, 1.8148167 \times 10^{-6}, 8.9890229 \times 10^{-7}, 4.4467752 \times 10^{-7}, 2.1968048 \times 10^{-7}, 1.0837093 \times 10^{-7}, 5.3379165 \times 10^{-8}, 2.6250177 \times 10^{-8}, 1.2887232 \times 10^{-8}, 6.3156953 \times 10^{-9}, 3.0894829 \times 10^{-9}, 1.5084277 \times 10^{-9}, 7.3503876 \times 10^{-10}, 3.5745232 \times 10^{-10}, 1.7347035 \times 10^{-10}, 8.4006106 \times 10^{-11}, 4.0593472 \times 10^{-11}, 1.9572421 \times 10^{-11}, 9.4158848 \times 10^{-12}, 4.5195275 \times 10^{-12}, 2.1643581 \times 10^{-12}, 1.0340940 \times 10^{-12}, 4.9292167 \times 10^{-13}, 2.3441025 \times 10^{-13}, 1.1121189 \times 10^{-13}, 5.2637877 \times 10^{-14}, 2.4855116 \times 10^{-14}, 1.1708532 \times 10^{-14}, 5.5024789 \times 10^{-15}, 2.5797915 \times 10^{-15}, 1.2066549 \times 10^{-15}, 5.6306241 \times 10^{-16}, 2.6212505 \times 10^{-16}, 1.2174281 \times 10^{-16}, 5.6411249 \times 10^{-17}, 2.6078415 \times 10^{-17}, 1.2028063 \times 10^{-17}, 5.5349743 \times 10^{-18}, 2.5412522 \times 10^{-18}, 1.1641243 \times 10^{-18}, 5.3208120 \times 10^{-19}, 2.4265609 \times 10^{-19}, 1.1041988 \times 10^{-19}, 5.0136548 \times 10^{-20}, 2.2715436 \times 10^{-20}, 1.0269655 \times 10^{-20}, 4.6330552 \times 10^{-21}, 2.0857623 \times 10^{-21}, 9.3703715 \times 10^{-22}, 4.2009978 \times 10^{-22}, 1.8795800 \times 10^{-22}, 8.3924933 \times 10^{-23}, 3.7398320 \times 10^{-23}, 1.6632335 \times 10^{-23}, 7.3825009 \times 10^{-24}, 3.2704840 \times 10^{-24}, 1.4460652 \times 10^{-24}, 6.3817537 \times 10^{-25}, 2.8111081 \times 10^{-25}, 1.2359739 \times 10^{-25}, 5.4243064 \times 10^{-26}, 2.3762443 \times 10^{-26}, 1.0391031 \times 10^{-26}, 4.5358179 \times 10^{-27}, 1.9764714 \times 10^{-27}, 8.5974956 \times 10^{-28}, 3.7334326 \times 10^{-28}, 1.6184823 \times 10^{-28}, 7.0045319 \times 10^{-29}, 3.0264255 \times 10^{-29}, 1.3054753 \times 10^{-29}, 5.6221577 \times 10^{-30}, 2.4173573 \times 10^{-30}, 1.0377414 \times 10^{-30}, 4.4478726 \times 10^{-31}, 1.9033751 \times 10^{-31}, 8.1312453 \times 10^{-32}, 3.4660513 \times 10^{-32}, 1.4712665 \times 10^{-32}, 6.1707325 \times 10^{-33}, 2.4790812 \times 10^{-33}, 8.2636039 \times 10^{-34})^t$,

using 40 iterations with tolerance 1.00×10^{-5} in l_∞ norm with the Gauss-Seidel method.

17. (a) Since A is a positive definite, $a_{ii} > 0$ for $1 \leq i \leq n$ and A is symmetric. Thus, A can be written as $A = D - L - L^t$, where D is diagonal with $d_{ii} > 0$ and L is lower triangular. The diagonal of the lower triangular matrix $D - L$ has the positive entries $d_{11} = a_{11}$, $d_{22} = a_{22}, \dots, d_{nn} = a_{nn}$, so $(D - L)^{-1}$ exists.

- (b) Since A is symmetric,

$$P^t = (A - T_g^t A T_g)^t = A^t - T_g^t A^t T_g = A - T_g^t A T_g = P.$$

Thus, P is symmetric.

- (c) $T_g = (D - L)^{-1} L^t$, so

$$(D - L)T_g = L^t = D - L - D + L + L^t = (D - L) - (D - L - L^t) = (D - L) - A.$$

Since $(D - L)^{-1}$ exists, we have $T_g = I - (D - L)^{-1} A$.

- (d) Since $Q = (D - L)^{-1} A$, we have $T_g = I - Q$. Note that Q^{-1} exists. By the definition of P we have

$$\begin{aligned} P &= A - T_g^t A T_g = A - [I - (D - L)^{-1} A]^t A [I - (D - L)^{-1} A] \\ &= A - [I - Q]^t A [I - Q] = A - (I - Q^t) A (I - Q) \\ &= A - (A - Q^t A) (I - Q) = A - (A - Q^t A - A Q + Q^t A Q) \\ &= Q^t A + A Q - Q^t A Q = Q^t [A + (Q^t)^{-1} A Q - A Q] \\ &= Q^t [A Q^{-1} + (Q^t)^{-1} A - A] Q. \end{aligned}$$

(e) Since

$$AQ^{-1} = A[A^{-1}(D - L)] = D - L \quad \text{and} \quad (Q^t)^{-1}A = D - L^t,$$

we have

$$AQ^{-1} + (Q^t)^{-1}A - A = D - L + D - L^t - (D - L - L^t) = D.$$

Thus,

$$P = Q^t [AQ^{-1} + (Q^t)^{-1}A - A] Q = Q^t D Q.$$

So for $\mathbf{x} \in \mathbb{R}^n$, we have $\mathbf{x}^t P \mathbf{x} = \mathbf{x}^t Q^t D Q \mathbf{x} = (Q\mathbf{x})^t D (Q\mathbf{x})$.

Since D is a positive diagonal matrix, $(Q\mathbf{x})^t D (Q\mathbf{x}) \geq 0$ unless $Q\mathbf{x} = \mathbf{0}$. However, Q is nonsingular, so $Q\mathbf{x} = \mathbf{0}$ if and only if $\mathbf{x} = \mathbf{0}$. Thus, P is positive definite.

(f) Let λ be an eigenvalue of T_g with the eigenvector $\mathbf{x} \neq \mathbf{0}$. Since $\mathbf{x}^t P \mathbf{x} > 0$,

$$\mathbf{x}^t [A - T_g^t A T_g] \mathbf{x} > 0$$

and

$$\mathbf{x}^t A \mathbf{x} - \mathbf{x}^t T_g^t A T_g \mathbf{x} > 0.$$

Since $T_g \mathbf{x} = \lambda \mathbf{x}$, we have $\mathbf{x}^t T_g^t = \lambda \mathbf{x}^t$ so

$$(1 - \lambda^2) \mathbf{x}^t A \mathbf{x} = \mathbf{x}^t A \mathbf{x} - \lambda^2 \mathbf{x}^t A \mathbf{x} > 0.$$

Since A is positive definite, $1 - \lambda^2 > 0$, and $\lambda^2 < 1$. Thus $|\lambda| < 1$.

(g) For any eigenvalue λ of T_g , we have $|\lambda| < 1$. This implies $\rho(T_g) < 1$ and T_g is convergent.