Eigenvalue Bounds for the Finite-State Birth-Death Process Intensity Matrix

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Introduction and Preliminaries



Continuous-Time Markov Chain

Definition (Continuous-Time Markov Chain)

A stochastic process $\{X(t): t \geq 0\}$ with a discrete state space S is a **continuous-time Markov chain** (CTMC) if for all $t \geq 0$, $s \geq 0$, and $i, j \in S$, it holds that

$$\mathbb{P}\left[X(t+s) = j|X(s) = i, \{X(u) : 0 \le u < s\}\right]$$

$$= \mathbb{P}\left[X(t+s) = j|X(s) = i\right].$$
(1)



Continuous-Time Markov Chain

- The quantity $P_{ij} = \mathbb{P}\left[X(t+s) = j | X(s) = i\right]$ is the probability of transitioning to state j from state i.
- For a CTMC, the future realization X(s + t) is only dependent on the current state X(s) and not on information occurring prior to time s.
- The matrix $P = [P_{ij}]$ corresponding to a CTMC is known as the matrix of **transition probabilities**.



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Birth-Death Process

Definition (Birth-Death Process)

A birth-death process is a continuous-time Markov chain for which

$$P_{ij} = \mathbb{P}\left[X(t+s) = j|X(s) = i\right] = 0$$

whenever |i - j| > 1.



Intensity Matrices

- Aside from transition probability matrices, information on CTMCs can also be stored in transition rate matrices or intensity matrices.
- The intensity matrix describes the rate at which a CTMC transitions or moves between states in its state space, thereby fully characterizing the process.
- The intensity matrix can also be used to determine the steady-state of the system.

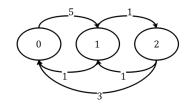


Intensity Matrices

Intensity matrix

$$\mathbf{Q} = \left[\begin{array}{rrr} -5 & 5 & 0 \\ 1 & -2 & 1 \\ 3 & 1 & -4 \end{array} \right]$$

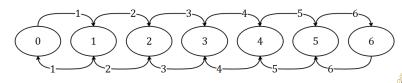
Diagram





Intensity Matrix of a Birth-Death Process

$$\mathbf{Q} = \begin{bmatrix} -1 & 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & -3 & 2 & 0 & 0 & 0 & 0 \\ 0 & 2 & -5 & 3 & 0 & 0 & 0 \\ 0 & 0 & 3 & -7 & 4 & 0 & 0 \\ 0 & 0 & 0 & 4 & -9 & 5 & 0 \\ 0 & 0 & 0 & 0 & 5 & -11 & 6 \\ 0 & 0 & 0 & 0 & 0 & 6 & -6 \end{bmatrix}$$





Intensity Matrices: Issues on Size

- Calculations involving the intensity matrix may prove inefficient as the state space becomes larger.
- Instead of directly dealing with the intensity matrix, we may investigate its eigenvalues.
- However, larger orders of the intensity matrix implies higher degrees of the characteristic polynomial for the eigenvalues whose roots may not be easily determined.



Statement of the Problem

- Determine upper and lower bounds for the eigenvalues of the intensity matrix of a finite-state birth-death process using principal sub-matrices of the intensity matrix.
- Examine how the bounds can be used to numerically compute for the eigenvalues of the given intensity matrix.



Existing Techniques on Eigenvalue Bounding



Bounds via Matrix Elements

- A bound using matrix order and matrix elements by Hirsch; [17]
- The Gerschgorin circle theorem; [13]
- The ovals of Cassini by Brauer; and [2]
- Another inequality giving regions for the eigenvalues by Ostrowski. [25]



Bounds via Matrix Norms

- A relation between eigenvalues and powers of the matrix by Householder; [19]
- An upper bound for the modulus of the largest eigenvalue by the same author; [19]
- A more precise bound by Lorch; and [22]
- Matrix partition norm inequality by Feingold. [8]



Bounds through Related Matrices

- One through the complex conjugate by Hirsch; [17]
- A result by Bodewig; and [1]
- Several theorems by Wittmeyer. [32]



Interlacing Property of Eigenvalues



Interlacing Sequences

Definition (Interlacing Sequences)

Consider two sequences of real numbers: $\lambda_1 \geq \cdots \geq \lambda_n$ and $\mu_1 \geq \cdots \geq \mu_m$ with m < n. The second sequence is said to interlace the first whenever

$$\lambda_i \ge \mu_i \ge \lambda_{n-m+i}, \qquad i = 1, \dots, m$$
 (2)

If m = n - 1, the interlacing inequalities become

$$\lambda_1 \ge \mu_1 \ge \lambda_2 \ge \mu_2 \ge \cdots \ge \mu_m \ge \lambda_n$$
.



Let Q be a transition rate matrix for the seven-state birth-death process:

$$\mathbf{Q} = \begin{bmatrix} -a_0 & \lambda_0 & 0 & 0 & 0 & 0 & 0 \\ \mu_1 & -a_1 & 0 & 0 & 0 & 0 & 0 \\ 0 & \mu_2 & -a_2 & \lambda_2 & 0 & 0 & 0 \\ 0 & 0 & \mu_3 & -a_3 & \lambda_3 & 0 & 0 \\ 0 & 0 & 0 & \mu_4 & -a_4 & \lambda_4 & 0 \\ 0 & 0 & 0 & 0 & \mu_5 & -a_5 & \lambda_5 \\ 0 & 0 & 0 & 0 & 0 & \mu_6 & -a_6 \end{bmatrix}$$
(3)

List the eigenvalues of \mathbf{Q} as $\mathcal{Q}=\{q_0,q_1,q_2,q_3,q_4,q_5,q_6\}$, an increasing sequence with distinct and negative elements.



Consider the principal submatrices

$$\mathbf{P} = \begin{bmatrix} -a_0 & \lambda_0 & 0 \\ \mu_1 & -a_1 & \lambda_1 \\ 0 & \mu_2 & -a_2 \end{bmatrix} \quad \text{and} \quad \mathbf{R} = \begin{bmatrix} -a_4 & \lambda_4 & 0 \\ \mu_5 & -a_5 & \lambda_5 \\ 0 & \mu_6 & -a_6 \end{bmatrix}$$

obtained upon omission of row and column four in \mathbf{Q} . Correspondingly, list the eigenvalues of \mathbf{P} and \mathbf{R} as increasing sequences $\mathcal{P} = \{p_0, p_1, p_2\}$ and $\mathcal{R} = \{r_4, r_5, r_6\}$, respectively.



Proposition

Given the preceding set-up and notation, the possible outcomes are as follows:

- **1** The sequence of eigenvalues $Q \setminus (P \cap R)$ from **Q** interlace with the sequence of eigenvalues $P \cup R$ from **P** and **R**.
- ② If $\mathcal{P} \cap \mathcal{R}$ is non-empty, then $\mathcal{P} \cap \mathcal{R} \subseteq \mathcal{Q}$. In other words, every eigenvalue of \mathbf{P} that is equal to an eigenvalue of \mathbf{R} is also an eigenvalue of \mathbf{Q} .



For the first item, assuming all are distinct, for example, and listing both sets of eigenvalues from P and R as a single increasing sequence

$$\mathcal{U} = \{u_0, u_1, u_2, u_3, u_4, u_5\},\$$

the interlacing property for the eigenvalues may be explicitly expressed as the inequality series,

$$q_0 \le u_0 \le q_1 \le u_1 \le q_2 \le u_2 \le q_3 \le u_3 \le q_4 \le u_4 \le q_5 \le u_5 \le q_6.$$



Outline of the Proof

- Consider a set of probabilities across states
- Write one as a sum of the convolutions of the others
- Take the Laplace transform of the outcome
- Rewrite some elements in the equation as entries of Q
- Further simplify the expression through eigenvalues of Q
- lacktriangle Acquire the eigenpolynomial of generator Q
- Examine cases and summarize results



We consider the following probabilities:

$$y_{3,3}(t) = \mathbb{P}(\{Q(s) = 3 \ \forall \ s \in [0, t] \mid Q(0) = 3\}$$
 (4)

$$y_{3,4}(t)dt = \mathbb{P}(\{Q(s) = 3 \ \forall \ s \in [0,t] \ \land Q(u) = 4$$

$$\exists \ u \in (t,t+dt) \mid Q(0) = 3\})$$
(5)

$$y_{3,2}(t)dt = \mathbb{P}(\{Q(s) = 3 \ \forall \ s \in [0,t] \ \land Q(u) = 2$$

$$\exists \ u \in (t,t+dt) \mid Q(0) = 3\})$$
 (6)

$$y_4(t)dt = \mathbb{P}(\{Q(s) \ge 4 \ \forall \ s \in [0, t] \ \land Q(u) = 3 \\ \exists \ u \in (t, t + dt) \mid Q(0) = 4\})$$
 (7)

$$y_2(t)dt = \mathbb{P}(\{Q(s) \le 2 \ \forall \ s \in [0, t] \ \land Q(u) = 3$$

 $\exists \ u \in (t, t + dt) \mid Q(0) = 2\})$



(8)

We can therefore write $x_{3,3}(t)$ as

$$x_{3,3}(t) = y_{3,3}(t) + y_{3,4}(t) * y_4(t) * x_{3,3}(t) + y_{3,2}(t) * y_2(t) * x_{3,3}(t)$$
(9)

where f(t) * g(t) represents the convolution of f(t) and g(t).

Taking the Laplace transform of both sides, we obtain

$$x_{3,3}^{\mathcal{L}}(s) = \frac{y_{3,3}^{\mathcal{L}}(s)}{1 - y_{3,4}^{\mathcal{L}}(s) \cdot y_4^{\mathcal{L}}(s) - y_{3,2}^{\mathcal{L}}(s) \cdot y_2^{\mathcal{L}}(s)}$$
(10)

where $h^{\mathcal{L}}(s)$ denotes the Laplace transform of h(t).



From Keilson [20], we obtain the following representations:

$$y_{3,3}^{\mathcal{L}}(s) = \frac{1}{s + \lambda_3 + \mu_3}$$
 $y_{3,4}^{\mathcal{L}}(s) = \frac{\lambda_3}{s + \lambda_3 + \mu_3}$ $y_{3,2}^{\mathcal{L}}(s) = \frac{\mu_3}{s + \lambda_3 + \mu_3}$

and

$$y_4^{\mathcal{L}}(s) = -\sum_{i=4}^6 \frac{c_i r_i}{s - r_i} \qquad y_2^{\mathcal{L}}(s) = -\sum_{i=0}^2 \frac{c_i p_i}{s - p_i},$$

where $c_i \ge 0$ for all i = 0, 1, ..., 6, $c_0 + c_1 + c_2 \le 1$, and $c_4 + c_5 + c_6 \le 1$.



From the representations made prior, we can thus write

$$x_{3,3}^{\mathcal{L}}(s) = \frac{\delta(s)}{\epsilon(s)},$$

where

$$\delta(s) = (s - p_0)(s - p_1)(s - p_2)(s - r_4)(s - r_5)(s - r_6)$$

$$\epsilon(s) = \delta(s)(s + \lambda_3 + \mu_3) + \mu_3 \sum_{i=0}^{2} \frac{c_i p_i \delta(s)}{s - p_i} + \lambda_3 \sum_{i=4}^{6} \frac{c_i r_i \delta(s)}{s - r_i}.$$
(11)

According to [20], $\epsilon(s)$ is the polynomial whose roots are the eigenvalues of the transition rate matrix **Q**.



CASE 1: If $p_i = r_j$ for some i = 0, 1, 2 and j = 4, 5, 6, $\epsilon(p_i) = \epsilon(r_j) = 0$, then $p_i = r_j$ is an eigenvalue of **Q**.

- Write the equal eigenvalues as q'_1, q'_2, \ldots, q'_k . Let the excess eigenvalues from both **P** and **R** be $b_1, b_2, \ldots, b_{6-k}$, in increasing order.
- An alternative form for $\epsilon(s)$ is attained:

$$\epsilon(s) = (s - q_1') \cdot (s - q_2') \cdots (s - q_{k-1}') \cdot (s - q_k') \cdot \epsilon_E(s).$$



- Since all eigenvalues from P and R are negative and $c_i > 0$ for all i, $(-1)^i \cdot \epsilon_E(b_i) > 0$ for i = 1, ..., 6 k.
- This means that the zeros of $\epsilon_E(s)$, the eigenvalues of **Q** that are not among q'_1, \ldots, q'_k , interlace with $b_1, b_2, \ldots, b_{k-1}$.



CASE 2: If all eigenvalues from P and R are distinct, then one has $b_1, b_2, b_3, b_4, b_5, b_6$, in sequence as well.

- Similar to what was justified earlier, $(-1)^i \cdot \epsilon(b_i) > 0$ for i = 1, ..., 6.
- This means that the roots of $\epsilon(s)$, the eigenvalues of \mathbf{Q} , interlace with b_1, b_2, \ldots, b_6 .



Extensions

- The same result also holds for the intensity matrix \mathbf{Q} of a finite-state birth-death process with n+1 states.
- One may also select any state q, 0 < q < n, around which the principal submatrices **P** and **R** are formed.
- The proof of the extended case follows the same arguments as the proof presented prior.



Numerical Approximation of Eigenvalues Using the Interlacing Property—An Example



Example

We consider the transition rate matrix for the M/M/1 queue with $\lambda_0 = \mu_0 = 2$, signifying equal arrival and service rates:

$$\mathbf{Q} = \begin{bmatrix} -2 & 2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 2 & -4 & 2 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 2 & -4 & 2 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & -4 & 2 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 2 & -4 & 2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 2 & -4 & 2 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 2 & -4 & 2 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 2 & -4 & 2 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 2 & -2 \end{bmatrix}$$



Example

We extract the 3×3 principal submatrix in the upper left corner and 5×5 principal submatrix in the lower right:

$$\mathbf{P} = \begin{bmatrix} -2 & 2 & 0 \\ 2 & -4 & 2 \\ 0 & 2 & -4 \end{bmatrix} \quad \text{and} \quad \mathbf{R} = \begin{bmatrix} -4 & 2 & 0 & 0 & 0 \\ 2 & -4 & 2 & 0 & 0 \\ 0 & 2 & -4 & 2 & 0 \\ 0 & 0 & 2 & -4 & 2 \\ 0 & 0 & 0 & 2 & -2 \end{bmatrix}.$$



Outline of the Implementation

- Implement a recursive strategy to bring forth the characteristic polynomial of the involved matrices;
- ② Solve for the eigenvalues of the the chosen 3×3 principal submatrix, preferably analytically;
- ullet (Quasi-recursion) Further extract two 2 \times 2 submatrices from the 5 \times 5 principal submatrix and compute for the eigenvalues;
- Numerically find the eigenvalues of the 5 \times 5 principal submatrix based on the pooled eigenvalues of the 2 \times 2 matrices; and
- $\ \, \ \,$ Follow a numerical method to find the eigenvalues of the 9×9 matrix through the completed set of bounds.



Numerically Solving the Characteristic Polynomial

- The characteristic polynomial of **Q** may be obtained following the routine of White [31] for tridiagonal matrices.
- In our study, we used the bisection method to numerically solve for the roots of the characteristic polynomial of Q as informed by the upper and lower bounds obtained from the interlacing property.



Concluding Remarks



Conclusion

- The proved interlacing property can be used to construct upper and lower bounds for each eigenvalue of a given intensity matrix.
- Given upper and lower bounds, numerical root-finding methods can then be performed more efficiently to approximate the actual eigenvalues of the intensity matrix.



Recommendations

Considerations for future work are:

- To identify the pattern in which the pooled eigenvalues interlace
- To attempt a different submatrix selection consider the case of overlapping submatrices
- To examine the implications of the theorem graphically
- To further investigate how the eigenvalue bounds can be used for calculations using intensity matrices (i.e. finding transition probabilities, steady states, etc.)
- To check for variations of the property for other Markov chains



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