

An experimental approach to the NIEP using algebraic geometry

Benjamin J. Clark
Washington State University
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1 Background

The NIEP and some subproblems

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NIEP

- Given a finite list $\Lambda = \{s_1, \dots, s_n\}$ of complex numbers, the nonnegative inverse eigenvalue problem (NIEP) asks for necessary and sufficient conditions such that Λ is the spectrum of an n -by- n nonnegative matrix.
- A different formulation of the NIEP asks instead for necessary and sufficient conditions for a list of real numbers to be the coefficients of the characteristic polynomial of a nonnegative matrix.

NIEP current status

- The NIEP is solved for $n \leq 4$.
- The solutions for $n = 1, 2, 3$ are fairly easy to derive from conditions involving the trace, reality, moment (sums of powers of the eigenvalues), and Perron condition (the largest eigenvalue in modulus is nonnegative).
- Meehan in 1998 and Torre-Mayo et al. in 2005 both solved the $n = 4$ case. Both of them are more complicated.

RNIEP

A simplification of the NIEP is the real NIEP (RNIEP) where we are given a list of real numbers and asked for necessary and sufficient conditions for it to be the spectra of a nonnegative matrix.

RNIEP current status

- This problem is also only solved for $n \leq 4$.
- The solution for $n = 4$ however is substantially easier, only requiring the trace to be nonnegative and for a Perron root to exist.
- The trace/Perron conditions are known not to be sufficient for $n \geq 5$.

SNIEP

A further simplification of the NIEP is the symmetric NIEP (SNIEP) where we are given a list of real numbers and asked for necessary and sufficient conditions for it to be the spectra of a symmetric nonnegative matrix.

SNIEP current status, more of the same

- This problem is also only solved for $n \leq 4$.
- The solutions for $n \leq 4$ are equivalent to the RNIEP.
- This raises the question: are the RNIEP and SNIEP different?

SNIEP and RNIEP are different

Egleston, Lenker, and Narayan showed that

$$\left(1, \frac{71}{97}, -\frac{44}{97}, -\frac{54}{97}, -\frac{70}{97}\right)$$

is realizable in the RNIEP, but not in the SNIEP.

Stochastic restrictions

- A final common restriction to these problems is to force all the matrices that are picked from to be either singly or doubly stochastic.
- For some problems like the NIEP adding the stochastic restriction and solving that problem would be equivalent to solving the original problems.
- For other problems like the SNIEP we can't pull from both stochastic and symmetric matrices at the same time and still get the same solution set.

DS-SNIEP

- When we add the doubly stochastic restriction to the SNIEP (DS-SNIEP) we get that matrices we are pulling being doubly stochastic symmetric.
- For the eigenvalues, this guarantees all eigenvalues real and in the interval $[-1, 1]$.
- This problem is solved only for $n \leq 3$.

Semialgebraic set

Definition

Let $\mathbb{F}$ be a real closed field. A subset S of $\mathbb{F}^n$ is called a semialgebraic set if it is a finite union of a finite intersection of sets defined by the solutions of polynomial inequalities of the form

$$\{x \in \mathbb{F}^n : P(x) * 0\}$$

where $*$ could be $=$, $\geq$, $\leq$, $>$, or $<$.

A semialgebraic set is called basic if no unions are needed.

Elementary symmetric functions

Definition

The k th elementary symmetric function of n complex numbers $\sigma = (\lambda_1, \lambda_2, \dots, \lambda_n)$, for $k \leq n$ is

$$S_k(\sigma) = \sum_{1 \leq i_1 < \dots < i_k \leq n} \prod_{j=1}^k \lambda_{i_j}.$$

Elementary symmetric functions example

For $n = 2$ the elementary symmetric functions are

$$S_1(\lambda_1, \lambda_2) = \lambda_1 + \lambda_2$$

$$S_2(\lambda_1, \lambda_2) = \lambda_1 \lambda_2$$

For $n = 3$ the elementary symmetric functions are

$$S_1(\lambda_1, \lambda_2, \lambda_3) = \lambda_1 + \lambda_2 + \lambda_3$$

$$S_2(\lambda_1, \lambda_2, \lambda_3) = \lambda_1 \lambda_2 + \lambda_1 \lambda_3 + \lambda_2 \lambda_3$$

$$S_3(\lambda_1, \lambda_2, \lambda_3) = \lambda_1 \lambda_2 \lambda_3$$

Sums of the principal minors

Definition

A minor of a matrix A is the determinant of sub matrix of A . A minor is called principal if the picked rows and columns of the submatrix are the same.

Definition

For a given matrix A , denote $E_k(A)$ to be the sums of the principal minors of size k of A .

Coefficients of the characteristic polynomial

For a given matrix A let

$$p_A(t) = t^n + (-1)a_1t^{n-1} + \cdots + (-1)^{n-1}a_{n-1}t + (-1)^na_n$$

be its characteristic polynomial, then

$$a_k = E_k(A) = S_k(\sigma(A)).$$

Embedding the NIEP as a semialgebraic set

Let $A \in M_n^+$ with spectra σ . Using the equality between $E_k(A)$ and $S_k(\sigma)$ we can embed nonnegative matrices and their spectra as a semialgebraic set in $\mathbb{R}^{n^2+n}$. The inequalities are

$$a_{ij} \geq 0 \text{ for all } i, j \in \{1, \dots, n\}$$

$$E_k(A) - S_k(\sigma) = 0 \text{ for all } k \in \{1, \dots, n\}$$

Question

What is wrong with the above semialgebraic set?

Embedding the NIEP cont.

- To overcome the fact that nonnegative matrices can have complex eigenvalues, we need to use the fact that the elementary symmetric functions maps complex conjugates to real numbers.
- Using this, we can make modified elementary symmetric functions based on the number of conjugate pairs.
- This gives that the solution to the NIEP is the projection of the union of $\lfloor (n-1)/2 \rfloor$ basic semialgebraic sets in $\mathbb{R}^{n^2+n}$

3x3 NIEP embedded semialgebraic set

If $\sigma = (\lambda_1, \lambda_2, \lambda_3)$ is real, then

$$a_{ij} \geq 0 \text{ for all } i, j \in \{1, 2, 3\}$$

$$E_1(A) - \lambda_1 - \lambda_2 - \lambda_3 = 0$$

$$E_2(A) - \lambda_1\lambda_2 - \lambda_1\lambda_3 - \lambda_2\lambda_3 = 0$$

$$E_3(A) - \lambda_1\lambda_2\lambda_3 = 0$$

or if $\sigma = (\lambda_1, \lambda_2 + i\lambda_3, \lambda_2 - i\lambda_3)$, then

$$a_{ij} \geq 0 \text{ for all } i, j \in \{1, 2, 3\}$$

$$E_1(A) - \lambda_1 - 2\lambda_2 = 0$$

$$E_2(A) - 2\lambda_1\lambda_2 - \lambda_2^2 - \lambda_3^2 = 0$$

$$E_3(A) - \lambda_1\lambda_2^2 - \lambda_1\lambda_3^2 = 0$$

SNIEP's and RNIEP's embedding as basic semialgebraic sets

For $A \in M_n^+$ with spectra σ , the SNIEP is embedded as

$$a_{ij} \geq 0 \text{ for all } i, j \in \{1, \dots, n\}$$

$$E_k(A) - S_k(\sigma) = 0 \text{ for all } k \in \{1, \dots, n\}$$

$$a_{ij} - a_{ji} = 0$$

The embedded SNIEP and RNIEP are basic semialgebraic sets in $\mathbb{R}^{n^2+n}$.

Project your way to a solution

- One of the most important properties of a semialgebraic set is that they are closed under projection. This is known as the Tarski–Seidenberg theorem.
- This means that if you can find a semialgebraic set in an embedded space, then we know there exists a semialgebraic set in the projected space.
- The reality condition and a finite union/intersection of polynomial inequalities solves the NIEP.

Cylindrical algebraic decomposition

- Collin's algorithm or Cylindrical algebraic decomposition is the algorithm for computing the projections of a semialgebraic set.
- It is a major improvement upon Tarski-Seidenberg's result by giving a straight forward approach to computing the projection.
- This means that the NIEP is solvable and the algorithm for solving it already exists. However...

Doubly exponential means computers don't help

- Cylindrical algebraic decomposition has a doubly exponential computing time.
- Computing the solution to the NIEP for $n = 2$ took less than 1 second.
- I stopped my attempt for computing the solution to the NIEP for $n = 3$ after 13 hours. The software claimed to be approximately 2% done.
- The first result that would be interesting would be, $n = 4$ and the first new result would be $n = 5$. So even with massive parallelization, computers won't be able to brute force the problem.

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Semialgebraic sets for the NIEP up to $n = 4$

Theorem (Torre-Mayo et. al. Theorem 3)

Let $P(x) = x^n + \sum_{j=1}^n (-1)^j c_j x^{n-j}$ be a characteristic polynomial of degree $n \geq 3$ of a nonnegative matrix A . Then

$$0 \leq c_1$$

$$0 \leq (n-1)c_1^2 - 2nc_2$$

$$c_3 \geq \begin{cases} \frac{n-2}{n} \left(-c_1 c_2 + \frac{n-1}{3n} \left(\left(c_1^2 - \frac{2n}{n-1} c_2 \right)^{3/2} + c_1^3 \right) \right) & \text{if } (n-1)(n-4)c_1^2 < 2(n-2)^2 c_2, \\ \frac{(n-1)(n-3)}{3(n-2)^2} c_1^3 - c_1 c_2 & \text{otherwise.} \end{cases}$$

Semialgebraic sets for the NIEP up to $n = 4$ cont.

In terms of the coefficients of the characteristic polynomial,

$$P(x) = x^n + \sum_{j=1}^n (-1)^j c_j x^{n-j},$$

the semialgebraic sets that make the NIEP are

- $n = 1$: $P(x) = x - c_1$

$$c_1 \geq 0$$

- $n = 2$: $P(x) = x^2 - c_1 x + c_2$

$$c_1 \geq 0 \wedge c_1^2 - 4c_2 \geq 0$$

Semialgebraic sets for the NIEP up to $n = 4$ cont.

- $n = 3$: $P(x) = x^3 - c_1x^2 + c_2x - c_3$

$$c_1 \geq 0 \wedge c_1^2 - 3c_2 \geq 0 \wedge 27c_3 + 9c_1c_2 - 2c_1^3 \geq 0 \wedge \\ -4c_1^3c_3 - c_1^2c_2^2 + 18c_1c_2c_3 + 4c_2^3 + 27c_3^2 \geq 0$$

- $n = 4$: $P(x) = x^4 - c_1x^3 + c_2x^2 - c_3x + c_4$

$$c_1 \geq 0 \wedge 4c_3 + 4c_1c_2 - c_1^3 \geq 0 \wedge 8c_3 + 2c_1c_2 - c_1^3 \geq 0 \wedge \\ (-c_2 \geq 0 \vee -27c_1^3c_3 - 9c_1^2c_2^2 + 108c_1c_2c_3 + 32c_2^3 + 108c_3^2 \geq 0)$$

Semialgebraic sets as a tool for the NIEP

Some things to note about the semialgebraic approach:

- The perron condition (The largest eigenvalue in modulus is positive) is not necessary.
- Solving the RNIEP using these methods also gives a solution to the NIEP.
- Checking where a given list is realizable can be done in linear time with respect to the number of polynomial inequalities.

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How to numerically build the boundary polynomials?

- We can think of the semialgebraic set for the coefficients of the characteristic polynomial as a n dimensional feasibility region defined by n^2 bounded and constrained parameters.
- With this, we can turn the problem into a series of optimization problems of the form,

$$\begin{aligned} \min E_j \quad \text{subject to} \\ E_i = c_i \text{ for } i \in \{1, \dots, j-1\} \end{aligned}$$

where the c_i values are within the feasibility region defined by $E_1, \dots, E_{j-1}$.

Approach applied to the DS-SNIEP

- As mentioned above, the DS-SNIEP is still open for $n \geq 4$.
- Since the matrices we are working with are stochastic and symmetric, we know that the perron root will be 1 and that the rest of the eigenvalues will be real. This allows us to plot the feasibility region in 3d.

Feasibility region 2d slices, E_1 dependence on E_2

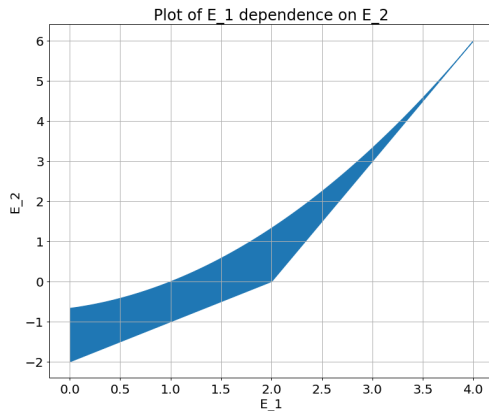


Figure 1: DS-SNIEP $n = 4$ E_1 dependence on E_2

Feasibility region 2d slices, E_1 dependence on E_3

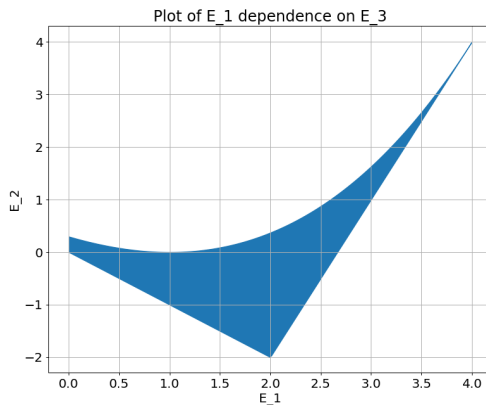


Figure 2: DS-SNIEP $n = 4$ E_1 dependence on E_3

Feasibility region 2d slices, E_2 dependence on E_3

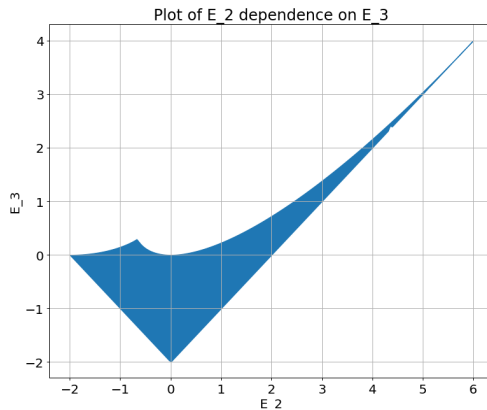
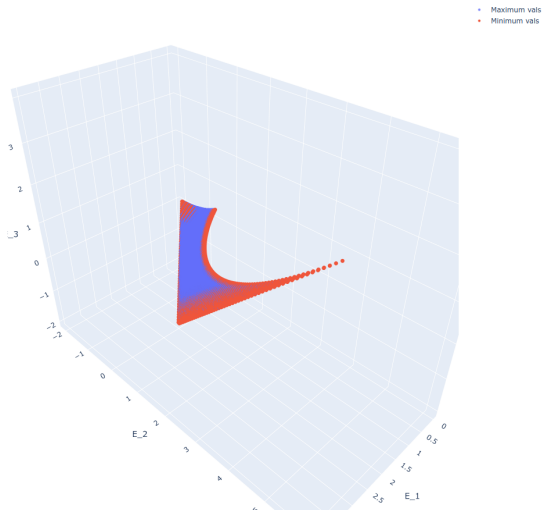


Figure 3: DS-SNIEP $n = 4$ E_2 dependence on E_3

Feasibility region of E_3

DS-SNIEP relation of E_1 , E_2 , and E_3 for $n=4$



Turn that region to the spectra

- The optimization process has the added benefit of returning the matrix that was found for each optimized point.
- Using this, we can turn this region of feasibility for the characteristic polynomials into a region of feasibility of the spectra.

Turn that region to the spectra

DS-SNIEP spectra region for $n=4$

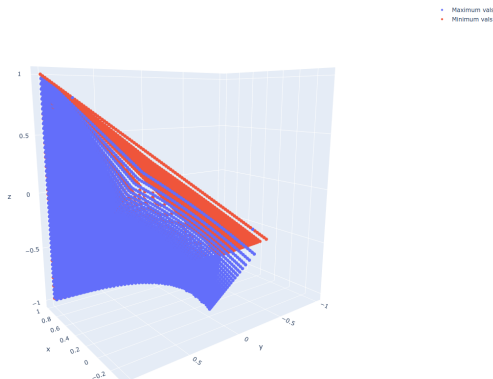


Figure 5: DS-SNIEP feasible spectral region

Bringing the numerical region to a conjecture

- Given the points on the boundary of a feasibility region. We can interpolate those points to build conjectures on solutions to these spectra problems.
- Two major drawbacks with this approach are the difficulty in finding the appropriate piecewise break points in higher dimensions and the numerical instability that arise from higher order optimizations.

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Theoretical improvements to the semialgebraic set.

- A major question with semialgebraic sets is how many polynomials and of what degree do we need them.
- The worst case of the projection operation is 2^n polynomials per variable projected and per polynomial in the initial constraints.
- My conjecture is that the NIEP forms a basic semialgebraic set with respect to its coefficients of the characteristic polynomial.
- This would reduce the worst case number of polynomials to n^2 .

Improvements to the solver

- Write code to determine where the break-lines are in the region.
- Allow for the solver to take more types of NIEP sub-problems in.
- Better interpolation support.

Interesting paper/book

- [1] Johnson, Marijuán, Paparella and Pisonero.
The NIEP.
Operator Theory, Operator Algebras, and Matrix Theory
- [2] Kamron Saniee.
A Simple Expression for Multivariate Lagrange Interpolation.
SIAM Undergraduate Research Online
- [3] Bochnak, Coste, and Roy.
Real Algebraic Geometry.
Springer-Verlag Berlin.

Questions?