# The Faddeev-LeVerrier algorithm

The Faddeev-LeVerrier algorithm may be used to determine the inverse, determinant, and characteristic polynomial of an  $n \times n$  matrix. The algorithm terminates in n steps, where each step involves a single matrix multiplication and only integer division. It works like magic!

FADDEEV-LEVERRIER ALGORITHM

1 **given** an  $n \times n$  matrix A

 $c_n \coloneqq 1$  $3 N \leftarrow 0$ 4 **for**  $k \in (n-1, ..., 1, 0)$  $5 \mid N \leftarrow N + c_{k+1} \mathbb{I}$ 

 $\begin{array}{c|c} \mathbf{8} & A^{-1} = -N/c_0 \\ \mathbf{9} & \det(A) = (-1)^n c_0 \\ \mathbf{10} & \chi(\lambda) = \sum_{k=0}^n c_k \lambda^k \end{array}$ 

Also refer to a Julia implementation.

### **Derivation**

Start with the characteristic polynomial of A.

$$\chi(\lambda) = \det(\lambda \mathbb{I} - A) = \sum_{k=0}^{n} c_k \lambda^k$$

## Useful fact.

The *adjunct* of a matrix,  $\operatorname{adj}(X)$ , satisfies  $\det(X)\mathbb{I} = A\operatorname{adj}(X)$ .

the entries of A. Hence, the entries of adj(X) are degree n-1 polynomials.

If A is  $n \times n$ , then  $\det(X)$ , and hence the entries of  $X \operatorname{adj}(X)$ , are degree n polynomials in

The entries of  $N(\lambda) := \operatorname{adj}(\lambda \mathbb{I} - A)$  are  $\lambda$ -polynomals of order n-1, so  $N(\lambda) = \sum_{k=0}^{n-1} N_k \lambda^k$ where  $N_k$  are matrices. From  $\det(\lambda \mathbb{I} - A)\mathbb{I} = (\lambda \mathbb{I} - A)N(\lambda)$ ,

$$\det(\lambda\mathbb{I}-A)\mathbb{I}=(\lambda\mathbb{I}-A)\sum_{k=0}^{n-1}N_k\lambda^k$$
 
$$=-AN_0+\sum_{k=1}^{n-1}(N_{k-1}-AN_k)\lambda^k+N_{n-1}\lambda^n$$
 Equating coefficients of  $\lambda$  with  $\chi(\lambda)\mathbb{I},$  we obtain:

 $c_0 \mathbb{I} = -AN_0$ 

$$c_0\mathbb{I}=N_{k-1}-AN_k$$
 
$$c_n\mathbb{I}=N_{n-1}$$
 To remember these, just write  $c_k\mathbb{I}=N_{k-1}-AN_k$  for all  $0\leq k\leq n$  with the understanding that

 $N_k$  vanishes outside the range  $0 \le k \le n-1$ . Equivalently,  $N_k = N_{k+1} - c_{k+1} \mathbb{I}$ 

gives a descending recurrence relation for 
$$N_k$$
 in terms of the coefficients  $c_k$ .

 $\mathcal{L}{f'(t)}(s) = \int_0^\infty f'(t)e^{-st} dt$ 

# This stroke of genious is due to (Hou, 1998).

Finding  $c_k$  in terms of A and  $N_k$ 

**Useful fact:** Laplace transform of derivative.

$$= f(t)e^{-st}\big|_{t=0}^{\infty} + s \int_{0}^{\infty} f(t)e^{-st} dt$$
$$= -f(0) + s\mathcal{L}\{f(t)\}(s)$$

Consider

and perform the Laplace transform to obtain 
$$-\mathbb{I}+s\mathcal{L}\big\{e^{At}\big\}=A\mathcal{L}\big\{e^{At}\big\}$$

 $\frac{\mathrm{d}e^{At}}{\mathrm{d}t} = Ae^{At}$ 

and finally take the trace:

 $\operatorname{diag}(J^k) = (\lambda_1^k, ..., \lambda_n^k).$ 

$$s\operatorname{tr}\mathcal{L}\!\left\{e^{At}\right\}-n=\operatorname{tr}\!\left(A\mathcal{L}\!\left\{e^{At}\right\}\right)$$

If  $\lambda_i$  are the eigenvalues of A then  $\operatorname{tr}(A) = \sum_i \lambda_i$ . Also, A can be put in Jordan normal form  $A = PJP^{-1}$  where J is triangular with  $\operatorname{diag}(J) = (\lambda_1, ..., \lambda_n)$ . Since it is triangular,

Therefore,  $\operatorname{tr}(A^k) = \operatorname{tr}(PJ^kP^{-1}) = \operatorname{tr}(J^kP^{-1}P) = \operatorname{tr}(J^k) = \sum_{i=1}^n \lambda_i^k$ . Consequently,  $\operatorname{tr}(e^{At}) = \sum_{k=0}^{\infty} \frac{t^k}{k!} \operatorname{tr}(A^k) = \sum_{k=0}^{\infty} \frac{t^k}{k!} \sum_{i=1}^n \lambda_i^k = \sum_{i=1}^n e^{\lambda_i t}$ .

We now compute the terms in Equation 1.

Useful fact: the trace of a matrix exponential in terms of eigenvalues.

$$\mathcal{L}\left\{e^{At}\right\} = \int_0^\infty e^{(A-s\mathbb{I})t} \, \mathrm{d}t = (A-s\mathbb{I})^{-1} e^{(A-s\mathbb{I})t} \Big|_{t=0}^\infty = (s\mathbb{I}-A)^{-1}$$

I'm uncomfortable with these indefinite integrals. Why should  $\lim_{t \to \infty} e^{(A-s\mathbb{I})t}$  converge?

Caution.

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

Note that from  $\chi(\lambda) = \det(\lambda \mathbb{I} - A) = (\lambda \mathbb{I} - A)N(\lambda)$  we have

 $(\lambda \mathbb{I} - A)^{-1} = \frac{N(\lambda)}{\gamma(\lambda)}$ 

 $\operatorname{tr} \mathcal{L} \big\{ e^{At} \big\} = \sum_{i=1}^n \frac{\mathrm{d}}{\mathrm{d}s} \ln(s - \lambda_i) = \frac{\mathrm{d}}{\mathrm{d}s} \ln \left( \prod_{i=1}^n (s - \lambda_i) \right) = \frac{\mathrm{d}}{\mathrm{d}s} \ln \chi(s) = \frac{\chi'(s)}{\chi(s)}$ 

 $\sum_{k=0}^{n} (k-n)c_k s^k = \sum_{k=0}^{n-1} \operatorname{tr}(AN_k)s^k$ 

 $\operatorname{tr} \mathcal{L} \{ e^{At} \} = \int_0^\infty \operatorname{tr} e^{(A-s\mathbb{I})t} dt = \sum_{i=1}^n \int_0^\infty e^{(\lambda_i - s)t} dt = \sum_{i=1}^n \frac{1}{s - \lambda_i}$ 

Let  $(\lambda_1,...,\lambda_n)$  be the eigenvalues of A. Then  $A-s\mathbb{I}$  has eigenvalues  $\lambda_i-s.$ 

Recall that the roots of the characteristic polynomial of 
$$A$$
 are its eigenvalues, so  $\chi(s)=\prod_{i=1}^n(s-\lambda_i).$ 

(1)

(2)

(3)

Substituting Equation 2 and Equation 3 into Equation 1, we have  $s\chi'(s) - n\chi(s) = \operatorname{tr}(AN(\lambda))$ 

$$c_k = \frac{\operatorname{tr}(AN_k)}{k-n}$$
 for all  $0 \le k \le n$  where we define  $N_n = 0$ .

 $\chi(\lambda) = \qquad \quad c_0 + \dots + \quad \ c_{n-1} \lambda^{n-1} + c_n \lambda^n$ 

 $c_0 = (-1)^n \det(A), \quad c_{n-1} = -\operatorname{tr}(A), \quad c_n = 1$ 

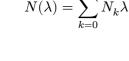
Useful fact.

which, expanding and equating powers of  $\lambda$ ,

Visual summary

Final algorithm

$$\det(\lambda \mathbb{I} - A) = \sum_{k=0}^{n} c_k \lambda^k \quad c_k \mathbb{I} = N_{k-1} - A N_k \quad c_k = \frac{\operatorname{tr}(A N_k)}{k-n} \qquad \mathcal{L}\{e^{At}\} = (s\mathbb{I} - A)^{-1}$$
 equate coefficients iterate 
$$c_0, \dots, c_n \\ N_0, \dots, N_{n-1} \qquad \qquad s \operatorname{tr} \mathcal{L}\{e^{At}\} - n = \operatorname{tr}(A \mathcal{L}\{e^{At}\})$$



References Hou, S.-H. (1998). Classroom Note: A Simple Proof of the Leverrier-Faddeev Characteristic

Polynomial Algorithm. SIAM Rev., 40(3), 706-709. https://doi.org/10.1137/S003614459732076X