

# 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

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1 given an  $n \times n$  matrix  $A$ 
2  $c_n := 1$ 
3  $N \leftarrow \mathbb{0}$ 
4  $k \leftarrow n - 1$ 
5 while  $k \geq 0$ 
6    $N \leftarrow N + c_{k+1}\mathbb{I}$ 
7    $c_k := \frac{1}{k-n} \operatorname{tr}(AN)$ 
8    $k \leftarrow k - 1$ 
9 return
10   $A^{-1} = -N/c_0$ 
11   $\det(A) = (-1)^n c_0$ 
12   $\chi(\lambda) = \sum_{k=0}^n c_k \lambda^k$ 
```

## 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  $\det(X)\mathbb{I} = A \operatorname{adj}(X)$ .

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  $A$ . Hence, the entries of  $\operatorname{adj}(X)$  are degree  $n - 1$  polynomials.

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

$$\begin{aligned} \det(\lambda \mathbb{I} - A)\mathbb{I} &= (\lambda \mathbb{I} - A) \sum_{k=0}^{n-1} N_k \lambda^k \\ &= -AN_0 + \sum_{k=1}^{n-1} (N_{k-1} - AN_k) \lambda^k + N_{n-1} \lambda^n \end{aligned}$$

Equating coefficients of  $\lambda$  with  $\chi(\lambda)\mathbb{I}$ , we obtain:

$$\begin{aligned} c_0 \mathbb{I} &= -AN_0 \\ c_k \mathbb{I} &= N_{k-1} - AN_k \\ c_n \mathbb{I} &= N_{n-1} \end{aligned}$$

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  $0 \leq k \leq n - 1$ . Equivalently,

$$N_k = N_{k+1} - c_{k+1} \mathbb{I}$$

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

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

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

### Useful fact: Laplace transform of derivative.

$$\begin{aligned} \mathcal{L}\{f'(t)\}(s) &= \int_0^\infty f'(t)e^{-st} \, dt \\ &= f(t)e^{-st} \Big|_{t=0}^\infty + s \int_0^\infty f(t)e^{-st} \, dt \\ &= -f(0) + s \mathcal{L}\{f(t)\}(s) \end{aligned}$$

Consider

$$\frac{d e^{At}}{dt} = A e^{At}$$

and perform the Laplace transform to obtain

$$-\mathbb{I} + s \mathcal{L}\{e^{At}\} = A \mathcal{L}\{e^{At}\}$$

and finally take the trace:

$$s \operatorname{tr} \mathcal{L}\{e^{At}\} - n = \operatorname{tr}(A \mathcal{L}\{e^{At}\}) \tag{1}$$

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

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  $\operatorname{diag}(J) = (\lambda_1, \dots, \lambda_n)$ . Since it is triangular,  $\operatorname{diag}(J^k) = (\lambda_1^k, \dots, \lambda_n^k)$ .

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.

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

### Caution.

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

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

$$(\lambda \mathbb{I} - A)^{-1} = \frac{N(\lambda)}{\chi(\lambda)} \tag{2}$$

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

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

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

$$\operatorname{tr} \mathcal{L}\{e^{At}\} = \sum_{i=1}^n \frac{d}{ds} \ln(s - \lambda_i) = \frac{d}{ds} \ln \left( \prod_{i=1}^n (s - \lambda_i) \right) = \frac{d}{ds} \ln \chi(s) = \frac{\chi'(s)}{\chi(s)} \tag{3}$$

Substituting Equation 2 and Equation 3 into Equation 1, we have

$$\begin{aligned} s\chi'(s) - n\chi(s) &= \operatorname{tr}(AN(\lambda)) \\ \sum_{k=0}^n (k-n)c_k s^k &= \sum_{k=0}^{n-1} \operatorname{tr}(AN_k) s^k \end{aligned}$$

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

$$c_k = \frac{\operatorname{tr}(AN_k)}{k - n}$$

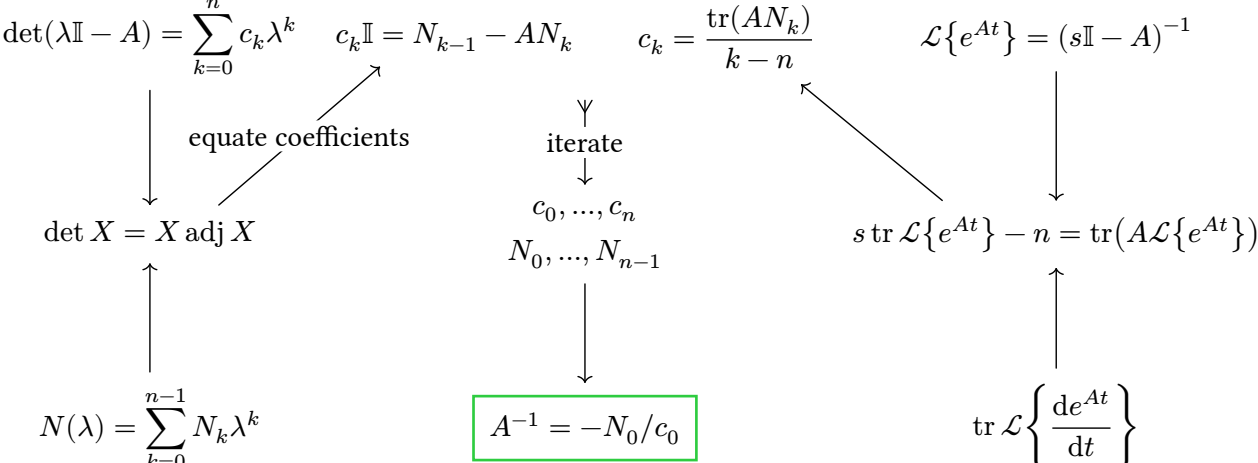
for all  $0 \leq k \leq n$  where we define  $N_n = 0$ .

## Final algorithm

### Useful fact.

$$\begin{aligned} \chi(\lambda) &= c_0 + \dots + c_{n-1} \lambda^{n-1} + c_n \lambda^n \\ &= \det(-A) + \dots + \operatorname{tr}(-A) \lambda^{n-1} + \lambda^n \\ c_0 &= (-1)^n \det(A), \quad c_{n-1} = -\operatorname{tr}(A), \quad c_n = 1 \end{aligned}$$

## Visual summary



## 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>