

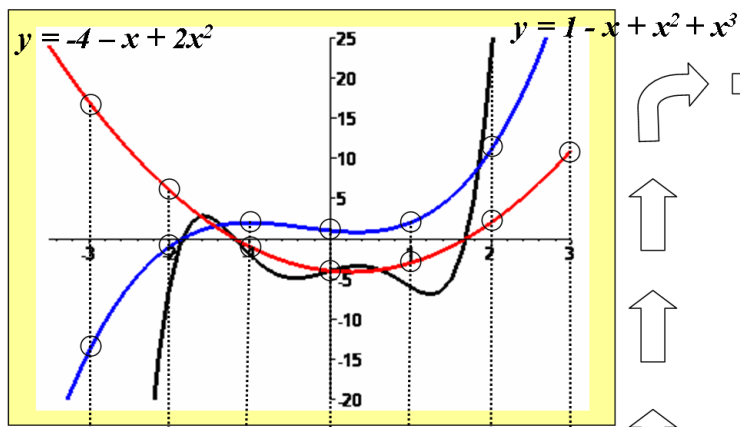


# THEOREM OF THE DAY

**The Convolution Theorem** Let  $a = (a_0, \dots, a_n)$  and  $b = (b_0, \dots, b_n)$  be vectors in  $\mathbb{C}^{n+1}$ . The convolution of  $a$  and  $b$ , denoted  $a \star b$ , is the vector  $c = (c_0, \dots, c_{2n})$ , in  $\mathbb{C}^{2n+1}$ , defined by  $c_i = \sum_{j=0}^n a_j b_{i-j}$ ,  $i = 0, \dots, 2n$ , with  $b_k = 0$  whenever  $k < 0$  or  $k > n$ . Then

$$a \star b = \mathcal{F}^{-1}(\mathcal{F}(a) \bullet \mathcal{F}(b)),$$

where  $\mathcal{F}$  is the  $(2n+1)$ -dimensional Discrete Fourier Transform and  $\bullet$  is componentwise multiplication.



	17	6	-1	-4	-3	2	11
$\bullet$	-14	-1	2	1	2	11	34
$=$	-238	-6	-2	-4	-6	22	374

$$\begin{aligned} c_0 + (-3)c_1 + (-3)^2c_2 + (-3)^3c_3 + (-3)^4c_4 + (-3)^5c_5 + (-3)^6c_6 &= -238 \\ c_0 + (-2)c_1 + (-2)^2c_2 + (-2)^3c_3 + (-2)^4c_4 + (-2)^5c_5 + (-2)^6c_6 &= -6 \\ c_0 + (-1)c_1 + (-1)^2c_2 + (-1)^3c_3 + (-1)^4c_4 + (-1)^5c_5 + (-1)^6c_6 &= -2 \\ c_0 + (0)c_1 + (0)^2c_2 + (0)^3c_3 + (0)^4c_4 + (0)^5c_5 + (0)^6c_6 &= -4 \\ c_0 + (1)c_1 + (1)^2c_2 + (1)^3c_3 + (1)^4c_4 + (1)^5c_5 + (1)^6c_6 &= -6 \\ c_0 + (2)c_1 + (2)^2c_2 + (2)^3c_3 + (2)^4c_4 + (2)^5c_5 + (2)^6c_6 &= 22 \\ c_0 + (3)c_1 + (3)^2c_2 + (3)^3c_3 + (3)^4c_4 + (3)^5c_5 + (3)^6c_6 &= 374 \end{aligned}$$

$$c_0 = -4, c_1 = 3, c_2 = -1, c_3 = -7, c_4 = 1, c_5 = 2, c_6 = 0$$

$$y = -4 + 3x - x^2 - 7x^3 + x^4 + 2x^5$$

The picture shows a version of the Convolution Theorem for polynomials:  $a \star b = \text{interpolate}(\text{evaluate}(a) \bullet \text{evaluate}(b))$ . 'Evaluate' means evaluate a degree  $n$  polynomial at  $2n+1$  points; the Discrete Fourier Transform (DFT) corresponds to a particular choice of points. Here we have merely chosen  $2n+1 = 7$  integers at which the two polynomials  $a : -4 - x + 2x^2$  and  $b : 1 - x + x^2 + x^3$  are evaluated (polynomial  $a$  is taken to have degree 3 but with  $a_3 = 0$ ). These values are multiplied and interpolation is applied: the 7 component products define a system of 7 simultaneous equations whose unique solution is precisely the list of coefficients of the product of the two polynomials (the coefficient of  $x^6$  is zero because polynomial  $a$  had degree  $< 3$ ). For polynomials, this is precisely what convolution achieves.

The DFT evaluates instead at complex numbers: the  $2n+1$  powers of a primitive  $(2n+1)$ -th root of unity. Via Euler's Identity, these are expressed as powers of  $e^{\tau i/(2n+1)}$  ( $\tau = 2\pi$ ). In general, to apply the DFT to a vector  $v$  of length  $N$ , we multiply  $v$  by the  $N \times N$  matrix  $\mathcal{F}$  whose  $ij$ -th element is  $\omega^{(i-1)(j-1) \pmod{N}}$ ,  $\omega = e^{\tau i/N}$ . (To our polynomial coefficient vectors we first need to append  $n$  zeros to extend to length  $2n+1$ .) This special choice of points leads to a dramatic computational short-cut: the so-called *Fast Fourier Transform* (FFT) achieves convolution (and hence polynomial multiplication) in time  $O(N \log N)$  instead of  $O(N^2)$ .

This discrete convolution theorem is intimately connected with the FFT known, in some form, to Gauss, as early as 1805; rediscovered by Cornelius Lanczos in 1940; and made widely known by James Cooley and John Tukey, 1965.

**Web link:** [betterexplained.com/articles/intuitive-convolution/](http://betterexplained.com/articles/intuitive-convolution/)

**Further reading:** *The Design and Analysis of Computer Algorithms* by A. Aho, J.E. Hopcroft and J.Ullman, Addison-Wesley, 1974, chapter 7.

