

## 9. POLYNOMIAL EQUATIONS

- Laguerre's method
- Sturm sequences
- Resultants

**Class web site:**

<http://www.di.ens.fr/~brette/calculscientifique/index.htm>

<http://www.di.ens.fr/~ponce/scicomp/notes.pdf>

## Newton's Method: Multivariate Case

What are the solutions of  $f_1(\vec{x}) = \dots = f_p(\vec{x}) = 0$ ?

- Taylor expansion of  $f_i$  in the neighborhood of  $x$ :

$$\begin{aligned} f_i(\vec{x} + \delta\vec{x}) &= f_i(\vec{x}) + \delta x_1 \frac{\partial f_i}{\partial x_1}(\vec{x}) + \dots + \delta x_q \frac{\partial f_i}{\partial x_q}(\vec{x}) + O(|\delta\vec{x}|^2) \\ &\approx f_i(\vec{x}) + \nabla f_i(\vec{x}) \cdot \delta\vec{x}, \end{aligned}$$

where  $\nabla f_i(\vec{x}) = (\partial f_i / \partial x_1, \dots, \partial f_i / \partial x_q)^T$  is the *gradient* of  $f_i$  at the point  $\vec{x}$

- This can be rewritten as

$$\vec{f}(\vec{x} + \delta\vec{x}) \approx \vec{f}(\vec{x}) + J_{\vec{f}}(\vec{x})\delta\vec{x} = \vec{0},$$

where  $J_{\vec{f}}(\vec{x})$  is the *Jacobian* of  $\vec{f} = (f_1, \dots, f_n)^T$ —that is, the  $p \times q$  matrix

$$J_{\vec{f}}(\vec{x}) \stackrel{\text{def}}{=} \begin{bmatrix} \nabla f_1^T(\vec{x}) \\ \dots \\ \nabla f_p^T(\vec{x}) \end{bmatrix} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1}(\vec{x}) & \dots & \frac{\partial f_1}{\partial x_q}(\vec{x}) \\ \dots & \dots & \dots \\ \frac{\partial f_p}{\partial x_1}(\vec{x}) & \dots & \frac{\partial f_p}{\partial x_q}(\vec{x}) \end{bmatrix}.$$

- **When  $p = q$ :** Iterate

$$\vec{x} = \vec{x} - J_{\vec{f}}^{-1}(\vec{x}) \vec{f}(\vec{x}).$$

- Quadratic convergence rate when it converges.

## Newton's Method for Nonlinear Least Squares

**When**  $p > q$ , define

$$E(\vec{x}) \stackrel{\text{def}}{=} \frac{1}{2} |\vec{f}(\vec{x})|^2 = \frac{1}{2} \sum_{i=1}^p f_i^2(\vec{x}),$$

and use Newton's method to find a local minimum of  $E$  as a zero of its gradient  $\vec{F}(\vec{x}) = \nabla E(\vec{x})$ .

- A simple calculation shows that

$$\vec{F}(\vec{x}) = J_{\vec{f}}^T(\vec{x}) \vec{f}(\vec{x}),$$

and differentiating this expression shows in turn that the Jacobian of  $\vec{F}$  is

$$J_{\vec{F}}(\vec{x}) = J_{\vec{f}}^T(\vec{x}) J_{\vec{f}}(\vec{x}) + \sum_{i=1}^p f_i(\vec{x}) H_{f_i}(\vec{x}),$$

where  $H_{f_i}(\vec{x})$  denotes the *Hessian* of  $f_i$ —that is, the  $q \times q$  matrix of second derivatives

$$H_{f_i}(\vec{x}) \stackrel{\text{def}}{=} \begin{bmatrix} \frac{\partial^2 f_i}{\partial x_1^2}(\vec{x}) & \cdots & \frac{\partial^2 f_i}{\partial x_1 \partial x_q}(\vec{x}) \\ \vdots & \ddots & \vdots \\ \frac{\partial^2 f_i}{\partial x_1 \partial x_q}(\vec{x}) & \cdots & \frac{\partial^2 f_i}{\partial x_q^2}(\vec{x}) \end{bmatrix}.$$

- The term  $\delta \vec{x}$  in Newton's method satisfies  $J_{\vec{F}}(\vec{x}) \delta \vec{x} = -\vec{F}(\vec{x})$ . Equivalently,  $\delta \vec{x}$  is the solution of

$$[J_{\vec{f}}^T(\vec{x}) J_{\vec{f}}(\vec{x}) + \sum_{i=1}^p f_i(\vec{x}) H_{f_i}(\vec{x})] \delta \vec{x} = -J_{\vec{f}}^T(\vec{x}) \vec{f}(\vec{x}).$$

## Variants of Newton's Method for Nonlinear Least Squares

**Gauss-Newton.** What is the value of  $\delta\vec{x}$  that minimizes  $E(\vec{x} + \delta\vec{x})$  for a given value of  $\vec{x}$ ?

$$E(\vec{x} + \delta\vec{x}) = |\vec{f}(\vec{x} + \delta\vec{x})|^2 \approx |\vec{f}(\vec{x}) + J_{\vec{f}}(\vec{x})\delta\vec{x}|^2.$$

- The adjustment  $\delta\vec{x}$  can be computed as the solution of  $J_{\vec{f}}^{\dagger}(\vec{x})\delta\vec{x} = -\vec{f}(\vec{x})$  or, equivalently, according to the definition of the pseudoinverse,

$$J_{\vec{f}}^T(\vec{x})J_{\vec{f}}(\vec{x})\delta\vec{x} = -J_{\vec{f}}^T(\vec{x})\vec{f}(\vec{x}).$$

- This is Newton where the Hessians  $H_{f_i}$  are taken equal to zero. This is justified when the residuals are small. Nearly quadratic convergence close to a solution.

### Levenberg-Marquardt.

- Take the increment  $\delta\vec{x}$  to be the solution of

$$[J_{\vec{f}}^T(\vec{x})J_{\vec{f}}(\vec{x}) + \mu\text{Id}]\delta\vec{x} = -J_{\vec{f}}^T(\vec{x})\vec{f}(\vec{x}),$$

where the parameter  $\mu$  is allowed to vary at each iteration.

- This is Newton where the term involving the Hessians is this time approximated by a multiple of the identity matrix. The Levenberg-Marquardt algorithm has convergence properties comparable to its Gauss-Newton cousin, but it is more robust.

## Real Algebraic Problems

*Example 1:* What are the real roots of  $x^4 - 3x^3 + x^2 - 5x + 1$ ?

*Example 2:* Do the surfaces defined in  $\mathbb{R}^3$  by  $x^2 - 5xy + 3z^3 = 0$  and  $-2x^3 + y^3 + 2xyz - 1 = 0$  intersect?

*Example 3:* If the intersection is not empty, how does it look?

*Example 4:* When is the ellipse defined by

$$\frac{(x - x_0)^2}{a^2} + \frac{(y - y_0)^2}{b^2} - 1 = 0$$

inside the unit circle centered at the origin?

Examples 1 and 3 are *display* problems, and example 2 is a *query* problem, and example 4 is a *constraint* problem.

Query and constraint problems can be reduced to *quantifier elimination* problems *over the reals*:

- Ex. 2:  $(\exists x)(\exists y)(\exists z)[x^2 - 5xy + 3z^3 = 0 \text{ and } -2x^3 + y^3 + 2xyz - 1 = 0]$ .
- Ex. 4:  $(\forall x)(\forall y)[b^2(x - x_0)^2 + a^2(y - y_0)^2 - a^2b^2 = 0 \implies x^2 + y^2 - 1 \leq 0]$ .

Display problems may reduce to root finding but may also boil down to determining the topology of a set of varieties. Quantifier elimination plays a role there too.

## Deflation

**Idea:** When a root  $\alpha$  of  $P(X)$  is found, we factor  $P(X)$  into

$$P(X) = (X - \alpha)Q(X),$$

where  $\deg Q(X) = \deg P(X) - 1$ . The roots of  $Q(X)$  are exactly the remaining roots of  $P(X)$ .

- A simple recurrence relation for deflation is easily found. Given

$$P(X) = a_n X^n + a_{n-1} X^{n-1} + \dots + a_1 X + a_0,$$

and

$$Q(X) = b_{n-1} X^{n-1} + \dots + b_0,$$

we clearly have:

$$\begin{cases} b_{n-1} = a_n, \\ b_{n-2} = a_{n-1} + \alpha b_{n-1}, \\ \dots \\ b_0 = a_1 + \alpha b_1. \end{cases}$$

- In floating-point arithmetic, deflation is not an exact process, hence error creeps in and accumulates, making the computation of successive roots of a polynomial less and less accurate.

- It is a good idea to treat the roots of the successively deflated polynomials as *tentative* roots of the original polynomial  $P(X)$ . They are *polished* by taking them as (good) initial guesses for the Newton method applied to the original *nondeflated* polynomial.

## Laguerre's Method

**Idea:** Design an iterative method guaranteed to converge to a root from any working point. Then use deflation to factor out this root, and iterate.

- Let  $P(X)$  be a polynomial of degree  $n$ , define

$$G(X) = \frac{P'}{P}(X) \quad \text{and} \quad H(X) = \left[\frac{P'}{P}(X)\right]^2 - \frac{P''}{P}(X).$$

- Set  $x$  to some initial value  $x_0$  and iterate  $x \leftarrow x - a$  where

$$a = \frac{n}{G(x) + s\sqrt{(n-1)(nH(x) - G^2(x))}},$$

until  $a$  is small enough. Here  $s = \mp 1$  is chosen at each iteration to maximize the magnitude of the denominator.

- **Note:** The computation of the radical requires complex arithmetic even when the coefficients of  $P(X)$  are real.

## Why Does It Work (Intuition)?

- Suppose that  $P(X)$  factorizes over  $\mathbb{C}$  as follows:

$$P(X) = (X - \alpha_1) \dots (X - \alpha_n).$$

It follows that

$$\frac{P'}{P}(X) = G(X) = \frac{1}{X - \alpha_1} + \dots + \frac{1}{X - \alpha_n},$$

and

$$\left[\frac{P'}{P}(X)\right]^2 - \frac{P''}{P}(X) = -G'(X) = H(X) = \frac{1}{(X - \alpha_1)^2} + \dots + \frac{1}{(X - \alpha_n)^2}.$$

- Now assume (*even though it is not true*) that the root  $\alpha_1$  that we seek is located at some distance  $a$  from our current guess  $x$ , while all other roots are assumed to be located at a distance  $b$ , i.e.,

$$a = x - \alpha_1, \quad b = x - \alpha_i, \quad \text{for } i = 2, \dots, n.$$

- This allows us to rewrite our constraints as

$$\begin{cases} \frac{1}{a} + \frac{n-1}{b} = G(x), \\ \frac{1}{a^2} + \frac{n-1}{b^2} = H(x). \end{cases}$$

- Solving this equation for  $a$  yields the Laguerre formulas.



## Euclid's Algorithm

**Theorem:** Let  $k$  be a commutative field, and let  $F(X)$  and  $G(X)$  be two nonzero polynomials in one variable over  $k$ . Then there exist unique polynomials  $Q(X), R(X)$  in  $k[X]$  such that  $F(X) = G(X)Q(X) + R(X)$  and  $\deg R(X) < \deg G(X)$ .

**Proof:** Write

$$F(X) = a_n X^n + \dots + a_1 X + a_0, \quad G(X) = b_d X^d + \dots + b_1 X + b_0,$$

where  $n = \deg F(X)$  and  $d = \deg G(X)$  so that  $a_n \neq 0$  and  $b_d \neq 0$ . We use induction on  $n$ .

If  $n = 0$  and  $\deg G(X) > \deg F(X)$ , we let  $Q(X) = 0$  and  $R(X) = F(X)$ . If  $\deg F(X) = \deg G(X) = 0$ , then we let  $R(X) = 0$  and  $Q(X) = a_n b_d^{-1}$ .

Now assume the theorem proved for  $n - 1$ . We may assume  $\deg G(X) \leq \deg F(X)$  (otherwise take  $Q(X) = 0$  and  $R(X) = F(X)$ ). Then we can write

$$F(X) = a_n b_d^{-1} X^{n-d} G(X) + F_1(X)$$

where  $F_1(X)$  has a degree strictly smaller than  $n$ . By the induction hypothesis, we can find  $Q_1(X), R(X)$  such that

$$F(X) = a_n b_d^{-1} X^{n-d} G(X) + Q_1(X)G(X) + R(X)$$

and  $\deg R(X) < \deg G(X)$ . Then we let

$$Q(X) = a_n b_d^{-1} X^{n-d} + Q_1(X)$$

which concludes the proof of existence of  $Q, R$ .

For uniqueness, suppose that

$$F(X) = Q_1(X)G(X) + R_1(X) = Q_2(X)G(X) + R_2(X)$$

with  $\deg R_1(X) < \deg G(X)$  and  $\deg R_2(X) < \deg G(X)$ . Subtracting yields

$$(Q_1(X) - Q_2(X))G(X) = R_2(X) - R_1(X),$$

so that

$$\deg (Q_1(X) - Q_2(X)) + \deg G(X) = \deg (R_2(X) - R_1(X)).$$

Since  $\deg (R_2(X) - R_1(X)) < \deg G(X)$ , this relation can only hold if  $Q_1(X) - Q_2(X) = 0$ , i.e.,  $Q_1(X) = Q_2(X)$  which in turns implies that  $R_1(X) = R_2(X)$ . QED.

## Squarefree Polynomials

• Consider a polynomial  $P(X)$  of degree  $n$  and its derivative  $P'(X)$ . Assuming that  $\alpha$  is a root of multiplicity  $m \leq n$  of  $P(X)$ , this polynomial can be written as

$$P(X) = (X - \alpha)^m Q(X),$$

where  $Q(X)$  is a polynomial of degree  $n - m$ . Hence the derivative of  $P(X)$  is

$$P'(X) = (X - \alpha)^{m-1} [mQ(X) + (X - \alpha)Q'(X)],$$

from which we deduce that  $\alpha$  is a root of multiplicity  $m - 1$  of  $P'$ , and that  $(X - \alpha)^{m-1}$  is a factor of the *greatest common divisor* (GCD) of  $P(X)$  and  $P'(X)$ .

• This leads to the following definition: The *squarefree part*  $P^*(X)$  of a polynomial  $P(X)$  is the quotient of  $P(X)$  by the greatest common divisor of  $P(X)$  and its derivative  $P'(X)$ .

• The polynomials  $P(X)$  and  $P^*(X)$  have exactly the same roots, but all the roots of  $P^*(X)$  are simple. If  $P(X) = P^*(X)$ , i.e.,  $P(X)$  has simple roots only, then  $P(X)$  is said to be *squarefree*.

## The Euclidean Algorithm

- Consider two integers  $x_0$  and  $x_1$ , with  $x_0 \geq x_1$ . We can define the following sequence of divisions:

$$\begin{aligned} x_0 &= a_1 x_1 + x_2, \\ x_1 &= a_2 x_2 + x_3, \\ &\dots \\ x_{n-1} &= a_n x_n + 0. \end{aligned}$$

The sequence stops when the last remainder is zero, at which point  $x_n$  is the GCD.

- Example: Computing the GCD of  $x_0 = 42$  and  $x_1 = 30$ . We have:

$$\begin{aligned} 40 &= 1 \times 30 + 12, \\ 30 &= 2 \times 12 + 6, \\ 12 &= 2 \times 6 + 0, \end{aligned}$$

and the GCD of 42 and 30 is 6.

- The same algorithm can be used to compute the GCD of two polynomials with rational coefficients.

## Exact Representation of Algebraic Numbers

- When a real  $\alpha$  is a root of some polynomial  $P(X)$  with rational coefficients, it is said to be an *algebraic number*. Some real numbers are not algebraic; e.g.,  $e$ , or  $\pi$  are *transcendental*.
- How can one represent *exactly* irrational algebraic numbers, such as  $\sqrt{2}$ ?
- An *isolating interval*  $[a, b]$  for  $\alpha$ , i.e., an interval bounded by rational numbers  $a < b$  that does not contain any other root of  $P(X)$ , together with the squarefree part  $P^*(X)$  of  $P(X)$ , provides an exact representation of  $\alpha$ .
- This representation also provides an arbitrarily precise numerical representation of  $\alpha$  through bisection.

## Sturm Sequences

• Let  $P(X)$  be a squarefree polynomial and  $P'(X)$  denote its derivative. The *Sturm sequence* of  $P(X)$  is the sequence  $\{F_i(X)\}$  of polynomials defined by

$$\begin{cases} F_0(X) = P(X), \\ F_1(X) = P'(X), \\ \text{for } i > 1, -F_i \text{ is the remainder obtained by dividing } F_{i-2} \text{ by } F_{i-1}. \end{cases}$$

Note: The degree of the polynomials  $F_i$  is strictly decreasing, ensuring that the sequence terminates in a finite number of steps with a constant polynomial  $F_k$ . This constant is different from zero since  $P(X)$  is squarefree.

• **Theorem:** For any interval  $[a, b]$  the number of roots of  $P(X)$  in  $[a, b]$  is  $S(a) - S(b)$ , where  $S(x)$  is the number of sign changes in the sequence  $[F_0(x), F_1(x), \dots, F_k(x)]$ .

Note: When  $a$  and  $b$  are rational numbers, we can compute the number of roots in  $[a, b]$  *exactly*.

• **Theorem:** Let  $P(X)$  be a polynomial of degree  $n$  with rational coefficients, the number of real zeros of  $P(X)$  that lie in any interval  $[a, b]$  with rational bounds can be found in  $O(n)$  arithmetic operations, after preprocessing that requires  $O(n \log^2 n)$  arithmetic operations.

## Root Isolation

- Given a squarefree polynomial  $P(X)$  with rational coefficients, first find an interval  $[a, b]$  that contains all its roots, e.g.,

$$b = -a = \max\left\{1 + \frac{p_i}{p_n} : i = 0, \dots, n-1\right\},$$

where  $p_i$  is the coefficient of the  $i$ th power of  $X$  in  $P(X)$ .

- Now, let  $N = S(a) - S(b)$  denote the number of real roots of  $P(X)$  in  $[a, b]$ . Use a binary search of  $I = [a, b]$  to find a point  $c$  in  $I$  which separates it into two subintervals, each containing at least one of the roots of  $P(X)$ . This can be done in  $O(n \log n)$  bisection steps, at each of which we must evaluate  $S$ , so that  $c$  can be found in  $O(n^2 \log n)$  time.
- Apply the same process to each of the subintervals  $[a, c]$  and  $[c, b]$  to find isolating intervals for all the roots of  $P(X)$ . At most  $N \leq n$  intervals will be processed, for a total cost of  $O(n^3 \log n)$  (which dominates the cost of the Sturm sequence computation).

## Example

- Consider the polynomial

$$P(X) = (X - 1)X(X + 1) = X^3 - X,$$

whose roots are  $-1, 0, 1$ .

- Its Sturm sequence is

$$\begin{cases} F_0(X) = P(X) = X^3 - X, \\ F_1(X) = P'(X) = 3X^2 - 1, \\ F_2(X) = 2/3X, \\ F_3(X) = 1. \end{cases}$$

Indeed, we have:

$$\begin{cases} F_0(X) = \frac{1}{3}X F_1(X) - F_2(X), \\ F_1(X) = \frac{9}{2}X F_2(X) - F_3(X). \end{cases}$$

- An interval  $[a, b]$  containing all roots is given by

$$b = -a = \max\left\{1 + \frac{0}{1}, 1 + \frac{-1}{1}, 1 + \frac{0}{1}, 1 + \frac{1}{1}\right\},$$

yielding the interval  $[-2, 2]$ .

- Next we compute  $S(-2)$  and  $S(2)$ . We have

$$\begin{cases} [F_0(-2) = -6, F_1(-2) = 11, F_2(-2) = -4/3, F_3(-2) = 1] \implies S(-2) = 3, \\ [F_0(+2) = +6, F_1(+2) = 11, F_2(+2) = +4/3, F_3(+2) = 1] \implies S(2) = 0, \end{cases}$$

which yields the (correct) number of roots in the interval  $[-2, 2]$ ,  $S(-2) - S(2) = 3$ .

- After that, the interval  $[-2, 2]$  can be subdivided until the three roots  $-1, 0, 1$  are isolated.

## Elimination Theory

**Theorem:** A necessary and sufficient condition for the system of  $n$  homogeneous linear equations in  $n$  unknowns

$$\begin{cases} a_{11}X_1 + \dots + a_{1n}X_n = 0, \\ \dots \\ a_{n1}X_1 + \dots + a_{nn}X_n = 0. \end{cases}$$

to admit a non-trivial solution is that

$$\text{Det} \begin{pmatrix} a_{11} & \dots & a_{1n} \\ & \dots & \\ a_{n1} & \dots & a_{nn} \end{pmatrix} = 0.$$

**Theorem:** A necessary and sufficient condition for the  $n$  homogeneous *polynomials* in  $\mathbb{Q}[X_1, \dots, X_n]$

$$\begin{cases} P_1(X_1, \dots, X_n) = 0, \\ \dots \\ P_n(X_1, \dots, X_n) = 0, \end{cases}$$

to admit a common root is that their *Macaulay resultant*  $R = \text{Det}(A)$  be equal to zero, where  $A$  is a matrix whose entries are polynomials in the coefficients of the polynomials  $P_i(X)$ .



## Sylvester Resultants

- Consider two polynomials  $P(X)$  and  $Q(X)$  in  $\mathbb{Q}[X]$  defined by

$$\begin{cases} P(X) = a_n X^n + \dots + a_1 X + a_0, \\ Q(X) = b_m X^m + \dots + b_1 X + b_0. \end{cases}$$

- Multiply  $P(X)$  successively by  $X^{m-1}, X^{m-2}, \dots, X$ , and 1, and multiply  $Q(X)$  by  $X^{n-1}, X^{n-2}, \dots, X$ , 1. This yields the following linear system in the power products  $X^{n+m-1}, X^{n+m-2}, \dots, X, 1$ :

$$\begin{pmatrix} a_n & a_{n-1} & \dots & a_1 & a_0 & & & \\ & a_n & a_{n-1} & \dots & a_1 & a_0 & & \\ \dots & & & & & & & \\ & & & a_n & a_{n-1} & \dots & a_1 & a_0 \\ b_m & b_{m-1} & \dots & b_1 & b_0 & & & \\ & b_m & b_{m-1} & \dots & b_1 & b_0 & & \\ \dots & & & & & & & \\ & & & b_m & b_{m-1} & \dots & b_1 & b_0 \end{pmatrix} \begin{pmatrix} X^{n+m-1} \\ \dots \\ X \\ 1 \end{pmatrix} = 0.$$

- Now if  $\alpha$  is a common root of the polynomials  $P(X)$  and  $Q(X)$ , then this system of homogeneous linear equations has a non-trivial solution, which implies in turn that the determinant of the  $(m+n) \times (m+n)$  matrix formed by the coefficients of  $P(X)$  and  $Q(X)$  is zero. This is the Sylvester resultant of these two polynomials.

- This also works for polynomials with coefficients in  $\mathbb{Q}[X_1, \dots, X_p] = \mathbb{Q}[X_1, \dots, X_{p-1}][X_p]$ .

## Bezout Resultants

- Consider two polynomials  $P_1, P_2$  of degree  $n$  in  $X$ , and the determinant:

$$D(X, Y) = \begin{vmatrix} P_1(X) & P_2(X) \\ P_1(Y) & P_2(Y) \end{vmatrix}.$$

- This determinant vanishes whenever  $X$  is a common root of  $P_1, P_2$  (the first row vanishes) and whenever  $X = Y$  (the two rows are identical). It follows that the polynomial  $D$  is divisible by  $(X - Y)$ , and that the polynomial:

$$F(X, Y) = D(X, Y)/(X - Y)$$

vanishes whenever  $X$  is a common root of  $P_1, P_2$ . Clearly,  $F$  has degree  $n - 1$  in  $X$  and  $Y$ . We can rewrite  $F$  as a polynomial in  $Y$ , so that:

$$F(X, Y) = \sum_{i=0}^{n-1} F_i(X)Y^i,$$

where  $F_i$  is a polynomial of degree  $n - 1$  in  $X$ .

## Bezout Resultants II

• Consider a common root  $\alpha$  of  $P_1(X)$  and  $P_2(X)$ . For any value  $Y$ , we have  $F(\alpha, Y) = 0$ , and it follows that all  $F_i(\alpha)$  are equal to zero. Let  $F_i^j$  denote the coefficient of degree  $j$  of  $F_i(X)$ , we obtain the following linear system:

$$\begin{pmatrix} F_0(\alpha) \\ \vdots \\ F_i(\alpha) \\ \vdots \\ F_{n-1}(\alpha) \end{pmatrix} = M \begin{pmatrix} \alpha^0 \\ \vdots \\ \alpha^j \\ \vdots \\ \alpha^{n-1} \end{pmatrix} = 0,$$

where:

$$M = \begin{pmatrix} F_0^0 & \dots & F_0^j & \dots & F_0^{n-1} \\ \vdots & & \vdots & & \vdots \\ F_i^0 & \dots & F_i^j & \dots & F_i^{n-1} \\ \vdots & & \vdots & & \vdots \\ F_{n-1}^0 & \dots & F_{n-1}^j & \dots & F_{n-1}^{n-1} \end{pmatrix}.$$

• Considering the successive powers  $\alpha^j$  as so many independent variables, it follows that this homogeneous linear system admits a non-trivial solution if and only if its determinant vanishes, i.e.  $\text{Det}(M) = 0$ . This determinant is Bezout's resultant.

• This generalizes easily into a method for eliminating two variables among three polynomials (*Dixon resultants*).

## Dixon Resultants

- Consider three polynomials  $P_1, P_2, P_3$  in two variables  $X_1, X_2$ , with highest degree  $n$  in  $X_1$  and  $m$  in  $X_2$ , and the determinant:

$$D(X_1, X_2, Y_1, Y_2) = \begin{vmatrix} P_1(X_1, X_2) & P_2(X_1, X_2) & P_3(X_1, X_2) \\ P_1(Y_1, X_2) & P_2(Y_1, X_2) & P_3(Y_1, X_2) \\ P_1(Y_1, Y_2) & P_2(Y_1, Y_2) & P_3(Y_1, Y_2) \end{vmatrix}.$$

- This determinant vanishes whenever  $(X_1, X_2)$  is a common root of  $P_1, P_2, P_3$  (the first row vanishes), and also whenever  $X_1 = Y_1$  (the first two rows are identical) or  $X_2 = Y_2$  (the last two rows are identical). It follows that the polynomial  $D$  is divisible by  $(X_1 - Y_1)(X_2 - Y_2)$ , and that the polynomial:

$$F(X_1, X_2, Y_1, Y_2) = D(X_1, X_2, Y_1, Y_2) / ((X_1 - X_2)(Y_1 - Y_2))$$

vanishes whenever  $(X_1, X_2)$  is a common root of  $P_1, P_2, P_3$ . Clearly,  $F$  has degree  $n - 1$  in  $X_1$ ,  $2m - 1$  in  $X_2$ ,  $2n - 1$  in  $Y_1$ , and  $m - 1$  in  $Y_2$ .

- We can rewrite  $F$  as a polynomial in  $Y_1$  and  $Y_2$ , so that:

$$F(X_1, X_2, Y_1, Y_2) = \sum_{\substack{i=0 \\ j=0}}^{\substack{2n-1 \\ m-1}} F_{i,j}(X_1, X_2) Y_1^i Y_2^j,$$

where  $F_{i,j}$  is a polynomial in  $X_1$  and  $X_2$ , with degree in  $X_1$  (resp.  $X_2$ ) less than or equal to  $n - 1$  (resp.  $2m - 1$ ).

## Dixon Resultants II

• Consider a common root  $(\alpha_1, \alpha_2)$  of  $P_1, P_2, P_3$ . For any value  $Y_1, Y_2$ , we have  $F(\alpha_1, \alpha_2, Y_1, Y_2) = 0$ , and it follows that all  $F_{i,j}(\alpha_1, \alpha_2)$  are equal to zero. Let  $F_{i,j}^{k,l}$  denote the coefficient of degree  $k$  in  $X_1$  and  $l$  in  $X_2$  of  $F_{i,j}$ , we obtain the following linear system:

$$\begin{pmatrix} F_{0,0}(\alpha_1, \alpha_2) \\ \vdots \\ F_{i,j}(\alpha_1, \alpha_2) \\ \vdots \\ F_{2n-1,m-1}(\alpha_1, \alpha_2) \end{pmatrix} = M \begin{pmatrix} \alpha_1^0 \alpha_2^0 \\ \vdots \\ \alpha_1^k \alpha_2^l \\ \vdots \\ \alpha_1^{n-1} \alpha_2^{2m-1} \end{pmatrix} = 0,$$

where

$$M = \begin{pmatrix} F_{0,0}^{0,0} & \cdots & F_{0,0}^{k,l} & \cdots & F_{0,0}^{n-1,2m-1} \\ \vdots & & \vdots & & \vdots \\ F_{i,j}^{0,0} & \cdots & F_{i,j}^{k,l} & \cdots & F_{i,j}^{n-1,2m-1} \\ \vdots & & \vdots & & \vdots \\ F_{2n-1,m-1}^{0,0} & \cdots & F_{2n-1,m-1}^{k,l} & \cdots & F_{2n-1,m-1}^{n-1,2m-1} \end{pmatrix}.$$

• Considering the successive powers  $\alpha_1^k \alpha_2^l$  as so many independent variables, it follows that this homogeneous linear system admits a non-trivial solution if and only if its determinant vanishes, i.e.,  $\text{Det}(M) = 0$ . This determinant is Dixon's resultant.