

# Notes for Lifting Techniques \*

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## 1 Preliminaries

We will be working with power series that have coefficients in  $\mathbb{Q}$ , denoted

$$\mathbb{Q}[[t]] = \left\{ \sum_{i \geq 0} c_i t^i \mid c_i \in \mathbb{Q} \right\}.$$

Let  $\mathbb{Q}[[t]]^{n \times n}$  be the  $n \times n$  matrices with entries from  $\mathbb{Q}[[t]]$ . Our development will be done in  $\mathbb{Q}[[t]]$  but our results will be valid for the more general case (i.e. replacing  $\mathbb{Q}$  with any ring).

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## 2 Introduction

Newton Iteration and Hensel lifting are iterative methods for finding a solution,  $x(t) \in \mathbb{Q}[[t]]$ , to some equation  $F(x(t), t) = 0$  where:

1.  $x(t)$  is a power series in  $t$ , i.e.  $x(t) = x_0 + x_1t + \dots$ .
2.  $x_0$  is known.

We would like to adapt this to do:

1. Inverse of power series, i.e. for  $1 - t \in \mathbb{Q}[[t]]$  calculate

$$(1 - t)^{-1} = 1 + t + t^2 + \dots$$

2. Inverse of matrices of power series, i.e. for

$$\mathbf{A} = \begin{bmatrix} \frac{1}{1-t} & 2+t \\ \frac{1}{1+t} & \frac{-3}{1+t^2+t^3} \end{bmatrix}$$

find  $\mathbf{A}^{-1} \in \mathbb{Q}[[t]]^{n \times n}$  such that  $\mathbf{A} \cdot \mathbf{A}^{-1} = \text{Id}$ .

3. Power series roots of univariate and multivariate equations, i.e.

$$y^2 - 1 - t = 0 \Rightarrow y = 1 + \frac{t}{2} - \frac{t^2}{2} + \dots$$

4. “Triangular sets” with power series coefficients.

We will study the development these methods.

## 3 Power Series Inversion

It is worth noting what constitutes an inverse of an element from  $\mathbb{Q}[[t]]$ . First year calculus teaches us that  $\frac{1}{1-t} = 1 + t + t^2 + \dots$  only when  $|t| < 1$  which can lead to some confusion. Reminding ourselves that the inverse of  $f$  (denoted  $f^{-1}$ ) uniquely satisfies  $ff^{-1} = 1$  we see that  $1 + t + t^2 + \dots$  is indeed the inverse of  $1 - t$ . Notice:

$$\begin{aligned} (1 - t)(1 + t + t^2 + \dots) &= (1 - t) + (1 - t)t + (1 - t)t^2 + \dots \\ &= 1 - t + t - t^2 + t^2 - t^3 + \dots \\ &= 1 \end{aligned}$$

We may also ask for the inverse of  $(1 - t) \in \mathbb{Q}[[t]]$  modulo  $t^k$ . In this case the inverse is  $1 + t + t^2 + \dots + t^{k-1}$  as:

$$\begin{aligned} (1 - t)(1 + t + t^2 + \dots + t^{k-1}) &\equiv 1 - t + t - t^2 + t^2 - t^3 + \dots - t^{k-1} + t^{k-1} + t^k \pmod{t^k} \\ &\equiv 1 \pmod{t^k} \end{aligned}$$

Our interest is devising algorithms that calculate these types of inverses up to some arbitrary  $k$  (usually a power of 2). In particular we are building an algorithm that has the following specification.

**Input** A series  $f(t) = f_0 + f_1t + f_2t^2 + \dots + f_nt^n \in \mathbb{Q}[[t]]$ ,  $f_0 \neq 0$  (otherwise  $f(t)$  has no inverse).

**Output**  $x(t) = x_0 + x_1t + x_2t^2 + \dots + x_nt^m \in \mathbb{Q}[[t]]$  such that  $x(t) \cdot f(t) = 1$ . (Note, we assume that  $\exists f_0^{-1}$  so  $x_0 = 1/f_0$ . This is a special case.)

### 3.1 By the Naive Algorithm

The basis of a naive algorithm is to extract the coefficients from  $xf = 1$  somehow. For  $a \in \mathbb{Q}[[t]]$  denote

$$[a]_i := \text{coefficient of } t^i \text{ in } a$$

so that  $[xf]_i = \sum_{j+k=i} x_j f_k$ . As  $xf = 1$  we have  $[xf]_i = \sum_{j+k=i} x_j f_k = 0$  for  $i > 0$ .

To develop the naive algorithm it is best to just work through an example.

**Example 1.** At each step we use  $[xf]_i$  to solve for  $x_i$  (note:  $1/f_0 = x_0$ );

$$\begin{aligned} i = 1 & \quad x_0 f_1 + x_1 f_0 = 0 & \Rightarrow x_1 = \frac{-x_0 f_1}{f_0} = -x_0 \\ i = 2 & \quad x_0 f_2 + x_1 f_1 + x_2 f_0 = 0 & \Rightarrow x_2 = \frac{-x_0 f_2 + x_1 f_1}{f_0} \\ i = 3 & \quad x_0 f_3 + x_1 f_2 + x_2 f_1 + x_3 f_0 = 0 & \Rightarrow x_3 = \frac{-x_0 f_3 + x_1 f_2 + x_2 f_1}{f_0} \end{aligned}$$

From Example 1 we see that we can calculate  $x_i$  by

$$x_i = \frac{-x_0 f_i + x_1 f_{i-1} + \cdots + x_{i-1} f_1}{f_0},$$

enabling us to generate the desired output. As we are not using information from  $x_{i-1}, \dots, x_0$  we do  $O(i)$  operations to get  $x_i$  for a total of  $O(i^2)$  operations to explicitly build  $x(t)$  to  $i$  terms which is far from ideal.

### 3.2 By Newton Iteration

We would like to reuse old information to save computation. So now suppose  $x_0, \dots, x_{i-1}$  ( $i$ -coefficients) in  $x(t)$  are given so that  $x(t)f(t) \equiv 1 \pmod{t^i}$ . Let

$$x(t) = x_0 + x_1 t + \cdots + x_{i-1} t^{i-1} + \delta x$$

where  $\delta x = a_i t^i + a_{i+1} t^{i+1} + \cdots$  are the higher order terms of  $x(t)$  whose coefficients are unknown. We can interpret this as knowing  $x(t) \pmod{t^i}$ . What follows is a method for establishing  $\delta x \pmod{t^{2i}}$  thereby allowing us to double the ‘‘accuracy’’ of  $x(t)$  (as we would expect from the quadratically convergent Newton’s method).

Define

$$x_{\text{init}} := x_0 + x_1 t + \cdots + x_{i-1} t^{i-1}$$

so that  $x = x_{\text{init}} + \delta x$ .

We again build coefficients by extracting them from  $xf = 1$  except now we have:

$$xf = 1 \Rightarrow (x_{\text{init}} + \delta x)f = 1 \Rightarrow x_{\text{init}}f + \delta x f = 1. \quad (1)$$

where (by our assumption)  $x_{\text{init}}f = 1 + 0t + \cdots + 0t^{i-1} + t^i R \equiv 1 \pmod{t^i}$  for some ‘‘remainder’’ term  $R$ . Multiplying (1) by  $x_{\text{init}}$  on both sides we get

$$x_{\text{init}}^2 f + x_{\text{init}} \delta x f = x_{\text{init}} \quad (2)$$

which allows us to derive an expression for  $\delta x$  as all other values are known.

Rewrite  $f x_{\text{init}} \equiv 1 \pmod{t^i}$  as  $x_{\text{init}} f = 1 + t^i R$  for some remainder  $R$  and multiply this expression by  $\delta x$  giving:

$$x_{\text{init}} \delta x f = \delta x + \delta x t^i R \quad (3)$$

Recall that  $\delta x \equiv 0 \pmod{t^i}$  so  $t^i | \delta x$  and therefore  $t^{2i} | \delta x t^i R$  meaning  $\delta x t^i R \equiv 0 \pmod{t^{2i}}$ . So, subbing (2) into (3) and taking  $\pmod{t^{2i}}$  we get

$$x_{\text{init}}^2 f + \delta x \equiv x_{\text{init}} \pmod{t^{2i}}$$

and solving for  $\delta x$  gives

$$\delta x \equiv x_{\text{init}} - x_{\text{init}}^2 f \pmod{t^{2i}} \quad (4)$$

which is the update formula we desire.

**Example 2.** By letting  $t = p$  for  $p$  some prime we can use this update formula to calculate inverses modulo  $p^n$ . If we let  $p = 3$  then we can calculate  $-1/2 \pmod{3^8 = 6561}$  as follows:

1.  $\frac{-1}{2} = \frac{1}{1-3} = 1 \pmod{3}$
2.  $\delta x \equiv (1 - (1)^2(1 - 3)) \pmod{3^2} = 3$  which implies  $1 + 3 = 4 \equiv \frac{-1}{2} \pmod{3^2}$
3.  $\delta x \equiv (4 - (4)^2(1 - 3)) \pmod{3^4} = 36$  which implies  $4 + 36 = 40 \equiv \frac{-1}{2} \pmod{3^4}$
4.  $\delta x \equiv (40 - (40)^2(1 - 3)) \pmod{3^8} = 3240$  which implies  $40 + 3240 = 3280 \equiv \frac{-1}{2} \pmod{3^8}$

where this process could be repeated up to any  $3^{2^k}$ .

To simplify the complexity analysis for this method we will assume that we can multiply polynomials in linear time (which is absurd as the best method is  $O(n \log n)$ ). Making this assumption means we will only be off by some log factors which is not a big deal.

Assuming that  $x_0 = 1/f_0$  is given it takes one operation to calculate  $x_1$ , two operations to calculate  $x_2, x_3$ , four operations to calculate  $x_4, \dots, x_7$ , and so on. Generalizing this we find that it takes  $O(1 + 2 + 4 + 8 + \dots + 2^k) = O(2^{k+1}) = O(2^k)$  operations to calculate  $O(2^k)$  terms.

**Remark 1.** An optimization to calculate  $x_{\text{init}}^2 f$  can be done. Observe

$$x_{\text{init}}^2 f = x_{\text{init}}(x_{\text{init}} f) = x_{\text{init}}(1 + 0t + \dots + 0t^{i-1} + t^i R) = x_{\text{init}} + t^i x_{\text{init}} R.$$

This means (3) can be rewritten as:

$$\delta x \equiv -t^i x_{\text{init}} R \pmod{t^{2i}}.$$

and using a trick called ‘‘middle product’’ it is possible to compute *only*  $R$  (see []).

## 4 Inversion of Matrices in $\mathbb{Q}[[t]]^{n \times n}$

Let  $\mathbf{F}(t) \in \mathbb{Q}[[t]]^{n \times n}$ , e.g. letting  $n = 2$  we have

$$\mathbf{F}(t) = \begin{bmatrix} f_{0,0}(t) & f_{0,1}(t) \\ f_{1,0}(t) & f_{1,1}(t) \end{bmatrix}$$

which we can express as a series of matrices (i.e. as an element from  $\mathbb{Q}^{2 \times 2}[[t]]$ ):

$$\mathbf{F}(t) = \mathbf{F}_0 + \mathbf{F}_1 t + \mathbf{F}_2 t^2 + \dots$$

where  $\mathbf{F}_i \in \mathbb{Q}^{2 \times 2}$ . What we would like to find is  $\mathbf{X}(t) = \mathbf{X}_0 + \mathbf{X}_1 t + \mathbf{X}_2 t^2 + \dots \in \mathbb{Q}^{n \times n}[[t]] \cong \mathbb{Q}[[t]]^{n \times n}$  such that  $\mathbf{F}\mathbf{X} = \text{Id}$ .

To do this:

1. Compute  $\mathbf{X}_0 = \mathbf{F}_0^{-1}$  (assume this is possible).
2. Repeat the newton iteration scheme from §3.2 replacing the series  $f(t), x(t)$  with the series of matrices  $\mathbf{F}(t), \mathbf{X}(t)$ . Namely update  $\mathbf{X} = \mathbf{X}_{\text{init}} + \delta\mathbf{X}$  using

$$\delta\mathbf{X} = \mathbf{X}_{\text{init}} - \mathbf{X}_{\text{init}}\mathbf{F}\mathbf{X}_{\text{init}} \pmod{t^{2i}}, \quad (5)$$

where the products are matrix multiplications.

**Remark 2.** The development of the above method can be done in the same manner as §3.2. Special care needs to be taken with regards to commutativity. However, it is true that

$$\mathbf{F}\mathbf{X}_{\text{init}} = \mathbf{X}_{\text{init}}\mathbf{F} \equiv 0 \pmod{t^i},$$

which is easily proved and useful for working out (5).

## 5 Series Roots of Univariate Polynomials

We now consider univariate polynomials with power series coefficients, i.e.  $F \in \mathbb{Q}[[t]][u]$  where

$$F(t, u) = u^2 - 1 - t - t^2 - t^3 - t^4 - \dots$$

Our goal is to compute a point  $x(t) \in \mathbb{Q}[[t]]$  such that  $F(t, x(t))|_{t=0} = 0$  (which will just write as  $F(0, x) = 0$ ). The point  $x = 1$  satisfies this property for  $F$  defined above.

For reasons that will become clear later we require

$$\frac{\partial F}{\partial u}(0, x) \neq 0.$$

We can interpret this geometrically as helping us avoid double roots (but more to the point we must eventually divide by this quantity).

For the algorithm assume we know  $x_0, x_1, \dots, x_{i-1}$  such that

$$F(t, x_0 + x_1 t + \dots + x_{i-1} t^{i-1}) \equiv 0 \pmod{t^i}.$$

We want to compute  $x_i$  such that

$$F(t, x_0 + x_1 t + \dots + x_i t^i) \equiv 0 \pmod{t^{i+1}} \quad (6)$$

**Definition 1** (Taylor formula). For a polynomial  $P$  we have

$$P(A + B) = P(A) + \frac{\partial P}{\partial u}(A)B + B^2 R \quad (7)$$

for  $R$  some polynomial remainder term.

Applying Taylor's formula to (6) with  $A = x_1 + \dots + x_{i-1}t^{i-1}$  and  $B = x_i t^i$  we get

$$0 \equiv F(A) + \frac{\partial F}{\partial u}(A)B + B^2 R \pmod{t^{i+1}} \quad (8)$$

$$\equiv F(t, x_0 + \dots + x_{i-1}t^{i-1}) + \frac{\partial F}{\partial u}(t, x_0 + \dots + x_{i-1}t^{i-1})x_i t^i + t^{2i} R \pmod{t^{i+1}} \quad (9)$$

The coefficient of  $t^i$  in (9) is

$$[F(t, x_0 + x_1 t + \dots + x_{i-1} t^{i-1})]_i + \left[ \frac{\partial F}{\partial u}(t, x_0 + \dots + x_{i-1} t^{i-1}) x_i t^i \right]_i$$

where

$$\left[ \frac{\partial F}{\partial u}(t, x_0 + \dots + x_{i-1} t^{i-1}) x_i t^i \right]_i = x_i \left[ \frac{\partial F}{\partial u}(t, x_0 + \dots + x_{i-1} t^{i-1}) \right]_0 = x_i \frac{\partial F}{\partial u}(0, x_0)$$

yielding the update formula

$$x_i = - \frac{[F(t, x_0 + \dots + x_{i-1} t^{i-1})]_i}{\frac{\partial F}{\partial u}(0, x_0)} \pmod{t^{i+1}}. \quad (10)$$

To instead lift a solution modulo  $t^i$  to modulo  $t^{2i}$  we apply the Taylor's formula to (9) using  $A = x_1 + \dots + x_{i-1}t^{i-1}$  and  $B = \delta x$ :

$$F(t, x_0 + \dots + x_{i-1}t^{i-1}) + \frac{\partial F}{\partial u}(t, x_0 + \dots + x_{i-1}t^{i-1})\delta x + \delta x^2 R. \quad (11)$$

Recall that  $\delta x \equiv 0 \pmod{t^i}$  and  $\delta x^2 \equiv 0 \pmod{t^{2i}}$  so taking (11) mod  $t^{2i}$  and solving for  $x_i$  gives:

$$\delta x = - \frac{F(t, x_0 + \dots + x_{i-1}t^{i-1})}{\frac{\partial F}{\partial u}(t, x_0 + \dots + x_{i-1}t^{i-1})} \pmod{t^{2i}}. \quad (12)$$

If we implement this we will have to:

1. compute  $F(t, x_0 + \dots + x_{i-1}t^{i-1}) \pmod{t^{2i}}$
2. compute  $\frac{\partial F}{\partial u}(t, x_0 + \dots + x_{i-1}t^{i-1}) \pmod{t^{2i}}$
3. invert and multiply  $\pmod{t^{2i}}$ .

However, we can reduce the complexity by some constant factors by making the following observation:

**Remark 3.** Since  $F(t, x_0 + \dots + x_{i-1}t^{i-1}) \equiv 0 \pmod{t^i}$  we may express it as  $t^i R_i$  with  $R_i \in \mathbb{Q}[[t]][u]$  and instead do:

$$\begin{aligned} &\equiv \frac{F(t, x_0 + \dots + x_{i-1}t^{i-1})}{\frac{\partial F}{\partial u}(t, x_0 + \dots + x_{i-1}t^{i-1})} \pmod{t^{2i}} \\ &\equiv \frac{t^i R_i}{\frac{\partial F}{\partial u}} \pmod{t^{2i}} \\ &= t^i \left( \frac{R_i}{\frac{\partial F}{\partial u}} \pmod{t^i} \right). \end{aligned}$$

Therefore we need only calculate  $\frac{\partial F}{\partial u}(t, x_0 + \dots + x_{i-1}t^{i-1}) \pmod{t^i}$  (instead of  $\pmod{t^{2i}}$ ).

**Remark 4** (Representation of  $F$ ).  $F$  is in  $k[[t]][u]$  so  $F = \sum_i F_i u^i$  for  $F_i \in k[[t]]$ . We need a data structure that can accommodate the evaluation of  $F$  (and its derivatives) at some arbitrary point. A DAG (directed acyclic graph) representation is a good choice. [PICTURE HERE]

## 6 Series Roots of Multivariate Polynomials

Let  $F_1, \dots, F_N$  be multivariate polynomials in  $\mathbb{Q}[[t]][u_1, \dots, u_n]$ . Our goal is to solve the system  $\langle F_1, \dots, F_n \rangle$  by finding  $x^{(1)}, \dots, x^{(n)} \in \mathbb{Q}[[t]]$  ( $x^{(i)} = x_0^{(i)} + x_1^{(i)}t + \dots$ ) such that

$$\begin{aligned} F_1(x^{(1)}, \dots, x^{(n)}, t)|_{t=0} &= 0 \\ &\vdots \\ F_n(x^{(1)}, \dots, x^{(n)}, t)|_{t=0} &= 0 \end{aligned}$$

We require a point  $(x_0^{(1)}, \dots, x_0^{(n)})$  satisfying

$$F_i(x_0^{(1)}, \dots, x_0^{(n)}) \text{ for } i = 1 \dots n$$

and that,

$$\mathbf{J} := \left[ \begin{array}{ccc} \frac{\partial F_1}{\partial u_1} & \dots & \frac{\partial F_1}{\partial u_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial F_n}{\partial u_1} & \dots & \frac{\partial F_n}{\partial u_n} \end{array} \right] \Big|_{(u_1, \dots, u_n) = (x^{(1)}, \dots, x^{(n)})}$$

is invertible  $\pmod{t}$  (i.e. the Jacobian of  $\mathbf{F} = \langle F_1, \dots, F_n \rangle$  evaluated at the initial point is invertible)

**Definition 2** (Generalized Taylor Formula). For a multivariate polynomial  $P$  we have

$$P(A^{(1)} + B^{(1)}, \dots, A^{(n)} + B^{(n)}) = P(A^{(1)}, \dots, A^{(n)}) + \sum \frac{\partial F}{\partial A^{(i)}} B^i + \langle B^{(1)}, \dots, B^{(n)} \rangle^2$$

Now, assume we have a solution for the system modulo  $t$ . Namely suppose that we are given

$$\begin{aligned} x^{(1)} &= x_0^{(1)} + x_1^{(1)}t + \dots + x_{i-1}^{(1)}t^i \\ &\vdots \\ x^{(n)} &= x_0^{(n)} + x_1^{(n)}t + \dots + x_{i-1}^{(n)}t^i \end{aligned}$$

such that

$$F_j(x^{(1)}, \dots, x^{(n)}) \equiv 0 \pmod{t^i} \text{ for } j = 1 \dots n.$$

To proceed with Newton iteration to find  $\delta x^{(1)}, \dots, \delta x^{(n)}$  such that

$$F_j(x^{(1)} + \delta x^{(1)}, \dots, x^{(n)} + \delta x^{(n)}) = 0 \text{ for } j = 1 \dots n.$$

apply Taylor's formula

$$F_j(x^{(1)}, \dots, x_n^{(n)}) + \sum_{k=1}^n \frac{\partial F_j}{\partial u_k}(x^{(1)}, \dots, x_n^{(n)}) \delta x^{(k)} + \langle \delta x^{(1)}, \dots, \delta x^{(n)} \rangle^2 = 0 \quad (13)$$

for  $j = 1 \dots n$  and reduce  $\pmod{t^{2i}}$  to get

$$F_j(x^{(1)}, \dots, x_n^{(n)}) + \sum_{k=1}^n \frac{\partial F_j}{\partial u_k}(x^{(1)}, \dots, x_n^{(n)}) \delta x^{(k)} + 0 \equiv 0 \pmod{t^{2i}} \quad (14)$$

$$\Rightarrow \left[ \frac{\partial F_j}{\partial u_1}(x^{(1)}, \dots, x_n^{(n)}), \dots, \frac{\partial F_j}{\partial u_n}(x^{(1)}, \dots, x_n^{(n)}) \right] \begin{bmatrix} \delta x^{(1)} \\ \vdots \\ \delta x^{(n)} \end{bmatrix} \equiv 0 \pmod{t^{2i}}$$

(note  $\delta x^{(j)} \delta x^{(j)} \equiv 0 \pmod{t^{2i}}$ ). This gives an expression for (13) in matrix form:

$$\mathbf{J} \begin{bmatrix} \delta x^{(1)} \\ \vdots \\ \delta x^{(n)} \end{bmatrix} \equiv - \begin{bmatrix} F_1(x^{(1)}, \dots, x_n^{(n)}) \\ \vdots \\ F_n(x^{(1)}, \dots, x_n^{(n)}) \end{bmatrix} \pmod{t^{2i}}$$

and solving gives an update formula for the  $\delta x^{(i)}$ 's:

$$\begin{bmatrix} \delta x^{(1)} \\ \vdots \\ \delta x^{(n)} \end{bmatrix} \equiv -\mathbf{J}^{-1} \begin{bmatrix} F_1(x^{(1)}, \dots, x_n^{(n)}) \\ \vdots \\ F_n(x^{(1)}, \dots, x_n^{(n)}) \end{bmatrix} \pmod{t^{2i}} \quad (15)$$

which has nontrivial implementation.

**Remark 5.** If we are in a lifting loop we reuse old  $\mathbf{J}^{-1}$ 's to update. Namely suppose that  $\mathbf{J} \pmod{t}$  is known, we compute  $\mathbf{J}^{-1} \pmod{t^{2i}}$  by computing  $\mathbf{J}^{-1} \pmod{t^2, t^4, \dots, t^{2i}}$  incrementally using lifting.

## 7 Lifting a Factor of a Univariate Polynomial

Let  $G(x, t), H(x, t), F(x, t) \in \mathbb{Q}[[t]][x]$ . Suppose  $G(x, t) \cdot H(x, t) \equiv F(x, t) \pmod{t^n}$ ,  $G, H \neq 1$  we call  $G$  and  $H$  the "factors" of  $F$  modulo  $t^n$

**Example 3.** Let

$$F(x, t) = x^4(1 + t + t^2 + t^3 + \dots) + 2x^3(1 + 4t + t^2 + \dots) \\ + x^2(3 + 3t + \dots) + 2x(1 + 4t + \dots) + (2 + 2t + \dots) + \dots$$

then  $G(x, t) = (x^2 + 1)(x^2 + 2x + 2)$  is a factor of  $F(x, 0)$ .



Assume a factor  $G_{\text{init}} = G_0 + tG_1 + \dots + t^{i-1}G_{i-1}$  of  $F \bmod t^n$  is known (we also require that  $\frac{\partial F}{\partial x}(x, 0)$  is invertible modulo  $G$  and that  $\deg_x G_k < \deg_x G_0$  for all  $k > 0$ ). We wish to find an update formula for  $\delta G$  so that

$$F = (G_{\text{init}} + \delta G)H. \quad (16)$$

(i.e., so that  $G = G_{\text{init}} + \delta G$  is a factor of  $F$  in the base field).

Now to get an update formula for  $\delta G$  recall  $\delta G \equiv 0 \bmod t^i$  and  $\frac{\partial \delta G}{\partial x} \equiv 0 \bmod t^i$ . This allows us to reduce

$$\frac{\partial G_{\text{init}}}{\partial x} F = \frac{\partial G_{\text{init}}}{\partial x} H (G_{\text{init}} + \delta G)$$

modulo  $t^{2i}$ , using

$$\begin{aligned} \frac{\partial F}{\partial x} &= \left( \frac{\partial G_{\text{init}}}{\partial x} + \frac{\partial \delta G}{\partial x} \right) H + (G_{\text{init}} + \delta G) \frac{\partial H}{\partial x} \\ \Rightarrow \frac{\partial G_{\text{init}}}{\partial x} H &= -\frac{\partial F}{\partial x} + \frac{\partial \delta G}{\partial x} H + (G_{\text{init}} + \delta G) \frac{\partial H}{\partial x} \end{aligned}$$

to get

$$\frac{\partial G_{\text{init}}}{\partial x} F \equiv \left( \frac{\partial G_{\text{init}}}{\partial x} H \right) G_{\text{init}} - \frac{\partial F}{\partial x} \delta G + \left( \frac{\partial H}{\partial x} \delta G \right) G_{\text{init}} \bmod t^{2i} \quad (17)$$

and taking this modulo  $G_{\text{init}}$  we get

$$\frac{\partial G_{\text{init}}}{\partial x} F \equiv -\frac{\partial F}{\partial x} \delta G \bmod \langle t^{2i}, G_{\text{init}} \rangle$$

yielding the update formula:

$$\delta G \equiv \left( -\frac{\partial G_{\text{init}}}{\partial x} \cdot F \right) / \left( \frac{\partial F}{\partial x} \right) \bmod \langle t^{2i}, G_{\text{init}} \rangle. \quad (18)$$

## 8 Lifting Triangular Sets

Lifting a triangular set can be interpreted as the generalization of lifting a root or factor of a polynomial.

We wish to devise an algorithm that has the following specification:

**Input** The system of polynomials  $F_1, \dots, F_n \in \mathbb{Q}[[t]][x_1, \dots, x_n]$  and triangular sets  $T_1, \dots, T_n$  with  $T_i \in \mathbb{Q}[x_i, \dots, x_n]$  such that

$$F_1(x_1, \dots, x_n, t) \equiv 0 \bmod \langle t, T_1, \dots, T_n \rangle$$

⋮

$$F_n(x_1, \dots, x_n, t) \equiv 0 \bmod \langle t, T_1, \dots, T_n \rangle$$

and

$$\begin{bmatrix} \frac{\partial F_1}{\partial x_1} & \dots & \frac{\partial F_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial F_n}{\partial x_1} & \dots & \frac{\partial F_n}{\partial x_n} \end{bmatrix}$$

is invertible  $\bmod \langle T_1(x_1, 0), \dots, T_n(x_1, \dots, x_n, 0) \rangle$ .

**Output** The triangular sets

$$\begin{aligned}
T_n(x_1, \dots, x_n, t) &= x_n^{d_n} + T_{n,d_n-1}(x_1, \dots, x_{n-1}, t)x_n^{d_n-1} + \dots + T_{n,0}(x_1, \dots, x_{n-1}, t) \\
&\vdots \\
T_2(x_1, x_2, t) &= x_2^{d_2} + T_{2,d_2-1}(x_1, t)x_2^{d_2-1} + \dots + T_{2,0}(x_1, t) \\
T_1(x_1, t) &= x_1^{d_1} + T_{1,d_1-1}(t)x_1^{d_1-1} + \dots + T_{1,0}(t)
\end{aligned}$$

such that  $F_i \equiv 0 \pmod{\langle t^N, T'_n, \dots, T'_1 \rangle}$ . That is, a “solution” in the sense that:

$$\begin{aligned}
&x_n - T_{N,0}(t) \\
&\vdots \\
&x_2 - T_{2,0}(t) \\
&x_1 - T_{1,0}(t).
\end{aligned}$$

It is *almost* the case that all roots of  $\{T_1, \dots, T_N\}$  are roots of  $\{F_1, \dots, F_n\}$  (almost because we may lose isolated roots).

**Example 4.** Let  $F_1, F_2 \in \mathbb{Q}[[t]][x_1, x_2]$  be given by:

$$\begin{aligned}
F_1 &= 1 + tx_1x_2 - t^2x_1 - (1+t)x_2 - x_1x_2^2 \\
F_2 &= t - (2t-1)x_1 + (1+t)x_1x_2 - tx_1^2x_2
\end{aligned}$$

corresponding to the triangular sets (our input)

$$\begin{aligned}
T_2(x_1, x_2, 0) &= x_1 + x_2 - 1 \\
T_1(x_1, 0) &= x_1^2 + 2x_1.
\end{aligned}$$

Since  $\begin{bmatrix} \frac{\partial F_1}{\partial x_1} & \frac{\partial F_1}{\partial x_2} \\ \frac{\partial F_2}{\partial x_1} & \frac{\partial F_2}{\partial x_2} \end{bmatrix}$  invertible mod  $\langle T_2(x_1, x_2, 0), T_1(x_1, 0), t \rangle$  our algorithm will output:

$$\begin{aligned}
T_2(x_1, x_2, t) &= x_2^2 + (1+t+\dots)x_1x_2 + (1+t+\dots)x_1 - (1+t+\dots) \\
T_1(x_1, t) &= x_1^2 + (1+t+\dots)x_1.
\end{aligned}$$

such that  $F_i \equiv 0 \pmod{\langle T_2, T_1, t^n \rangle}$ .

**Theorem 1.** Let  $F_1, \dots, F_n \in \mathbb{Q}[[t]][x_1, \dots, x_n]$ . The number of solutions of  $\langle F_1, \dots, F_n \rangle$  is bounded by  $\prod_{i=1}^n \deg(F_i)$

*Proof.* ? □

Let us now begin to develop the desired algorithm. First recall that  $F_i$  is required to reduce to 0 mod  $\langle T_1, \dots, T_n \rangle$ , we claim this is if and only if there exists  $H_{i,1}, \dots, H_{i,n}$  such that

$$F_i = H_{i,1}T_1 + \dots + H_{i,n}T_n.$$

(That is,  $F_i$  is a linear combination of elements in  $\{T_1, \dots, T_n\}$ .)

We can write the requirement that  $F_1, \dots, F_n$  reduces to 0 mod  $\langle T_1, \dots, T_n \rangle$  as a matrix expression:

$$\begin{bmatrix} F_1 \\ \vdots \\ F_n \end{bmatrix} = \begin{bmatrix} H_{1,1} & \cdots & H_{1,n} \\ \vdots & \ddots & \vdots \\ H_{n,1} & \cdots & H_{n,n} \end{bmatrix} \begin{bmatrix} T_1 \\ \vdots \\ T_n \end{bmatrix}$$

(note the  $H_{i,j}$ 's aren't necessarily unique, which is problematic).

Suppose we know  $(T_{1,\text{init}}, \dots, T_{n,\text{init}})$  such that  $F_1, \dots, F_n$  reduces to 0 mod  $\langle t^i, T_{1,\text{init}}, \dots, T_{n,\text{init}} \rangle$ . We would like to find  $\delta_1, \dots, \delta_n$  (polynomials) such that  $\delta_1 \in \mathbb{Q}[[t]][x_1]$ ,  $\delta_2 \in \mathbb{Q}[[t]][x_1, x_2]$ , and so on, such that

$$\begin{bmatrix} \delta_1 \\ \vdots \\ \delta_n \end{bmatrix} \equiv \begin{bmatrix} \frac{\partial T_1}{\partial x_1} & \cdots & \frac{\partial T_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial T_n}{\partial x_1} & \cdots & \frac{\partial T_n}{\partial x_n} \end{bmatrix} \begin{bmatrix} \frac{\partial F_1}{\partial x_1} & \cdots & \frac{\partial F_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial F_n}{\partial x_1} & \cdots & \frac{\partial F_n}{\partial x_n} \end{bmatrix}^{-1} \begin{bmatrix} F_1 \\ \vdots \\ F_n \end{bmatrix} \pmod{\langle t^{2i}, T_{1,\text{init}}, \dots, T_{n,\text{init}} \rangle}$$

(each one of these steps is non-trivial).

We will omit the proof of this development as it is similar enough to the proof given in §7.