

demoralizer's blog

[Tutorial] Solving Linear Recurrences with various methods, Including $O(N \log N \log K)$ using FFT

By [demoralizer](#), [history](#), 17 months ago, 

Hi this is my first educational blog on codeforces.....

I have been procrastinating to write this one for over 15 months by now, but thanks to [crackersamdjam](#)'s [comment](#), I finally did it!

Any suggestions or improvements or constructive-criticism to the blog, is heavily appreciated.

In this blog, I tried to avoid too many technical terms (especially in the last section), and tried to make it beginner friendly.

Linear Recurrence : Introduction

Fibonacci Sequence is one of the simplest examples of a linear recurrence:

$$F_x = F_{x-1} + F_{x-2}, \text{ with } F_0 = 0, F_1 = 1$$

Here is a more general example of a linear recurrence:

$$R_x = \sum_{i=1}^n C_i \cdot R_{x-i}$$

where C_i is constant and R_x is the x -th term of the recurrence. Since R_x depends on the previous n terms of the recurrence, it is called a linear recurrence of order n . It is called a "Linear" Recurrence, since we do not have any terms of the type $R_x \cdot R_y$

Note: We need n initial conditions for a linear recurrence of order n , for example, we have 2 initial conditions for the Fibonacci Sequence.

Iternary

In this blog we will learn about various methods of solving the following problem: Given a Linear Recurrence of order N , find the K -th term of the recurrence. There are various methods to do this:

- $O(N \cdot K)$ using DP
- $O(N^3 \cdot \log K)$ using Matrix Exponentiation
- $O(P(n) \cdot \log K)$ using Characterstic Polynomials where $P(n)$ is the time required for polynomial multiplication — which can be $O(N^2)$ naively or $O(N^{1.58})$ using Karatsuba Multiplication or $O(N \log N)$ using Fast Fourier Transform.

DP Solution

The $O(N \cdot K)$ solution is pretty trivial. (Assume `dp[0]` .. `dp[n-1]` are stored correctly as initial conditions)

```
for(int i = n; i < k; i++){
    dp[i] = 0;
    for(int j = 1; j <= n; j++){
        dp[i] += c[j] * dp[i - j];
    }
}
```

Matrix Exponentiation

You can find more detailed explanations on this technique [here](#) and [here](#). But here I've described it briefly:

Let's look at the problem from another angle. From the DP solution, it is clear that we need to keep track of the previous n elements at all times, and it won't matter even if we "forget" other elements. So let's keep a column vector of the "current" n consecutive elements. Since the recurrence relation is a Linear Relation, it is possible to have a linear transformation to get the next elements.

$$\begin{bmatrix} R_{n-1} \\ R_{n-2} \\ \vdots \\ R_0 \end{bmatrix} \xrightarrow{\text{A Linear Transformation}} \begin{bmatrix} R_n \\ R_{n-1} \\ \vdots \\ R_1 \end{bmatrix}$$

Finding this linear transformation is quite intuitive and it can be expressed as multiplication with a square matrix, so let's directly write the general result.

$$\begin{bmatrix} C_1 & C_2 & \cdots & C_{n-1} & C_n \\ 1 & 0 & \cdots & 0 & 0 \\ 0 & 1 & \cdots & 0 & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \cdots & 1 & 0 \end{bmatrix} \begin{bmatrix} R_{i+(n-1)} \\ R_{i+(n-2)} \\ R_{i+(n-3)} \\ \vdots \\ R_{i+(0)} \end{bmatrix} \implies \begin{bmatrix} R_{i+(n)} \\ R_{i+(n-1)} \\ R_{i+(n-2)} \\ \vdots \\ R_{i+(1)} \end{bmatrix}$$

Reader is advised to take a moment to manually multiply the matrices and verify the above result for a better understanding.

Alright, so a multiplication with the Transformation Matrix shifts the elements of the vector by 1 index each, so we can shift it by X indices by multiplying the transformation matrix to it X times. However we take advantage of the fact that we can first multiply the transformation with itself X times and then multiply it with the column vector. Why is this advantageous? Because we can use Binary Exponentiation! If we the transformation matrix is T , we can find T^X in $O(N^3 \log X)$ time. This lets us get the K -th term in $O(N^3 \log K)$ time.

[Check out this Mashup for practice problems](#)

Using Characterstic Polynomial

I might've over-explained this section of the blog, because I personally found it very hard to understand this part when I initially learnt it.

*I learnt about this technique from **TLE's** [blog on Berlekamp Massey Algorithm](#), but since it wasn't explained in detail, I had a lot of difficulty in understanding this, so I'll try to explain it in a more beginner-friendly way. Apparently the name of this trick is [Kitamasa Method](#).*

First of all, an important observation

The original recurrence can be re-written as:

$$R_j - \sum_{i=1}^n C_i \cdot R_{j-i} = 0$$

Which means, any multiple of $(R_j - \sum_{i=1}^n C_i \cdot R_{j-i})$ is 0 and if it is added or subtracted anywhere, the answer doesn't change.

Let's bring in Polynomials

In simple words, in the linear recurrence, replace R_i with x^i everywhere to get a polynomial equation. Multiplying x to every element of the polynomial would just mean adding 1 to all the

subscripts in the linear recurrence. It is quite obvious to see that adding or subtracting such polynomial equations is also allowed. *"Allowed? For what?"* — **For converting the polynomial back to the recurrence!** Yes! This is a bit magical.

Try to check it on a few cases manually to verify that it is always allowed to do the following:

- Convert the linear recurrence into polynomial equation
- Add/Subtract/Multiply the polynomial equation with x or some constants
- Convert the polynomial back to linear recurrence (replace x^i with R_i everywhere)

The final equation that you get after this, will be valid.

Alright, so what now? Let's convert the original recurrence into a polynomial (It is called the **Characteristic Polynomial** of the Recurrence)

$$f(x) = x^n - \sum_{i=1}^n C_i \cdot x^{n-i}$$

Now since we know that upon converting the characteristic polynomial back to recurrence form, its value will be 0. We take advantage of this — Any multiple of the characteristic polynomial can be added or subtracted without any effect on the result.

We want to find R_K , upon converting it to polynomial we get x^K

We wish to find $(x^K + m(x) \cdot f(x))$, where $m(x)$ is any polynomial! (Since $f(x)$ will be 0 upon converting back to recurrence)

Now the idea is simple, we choose $m(x)$ in such a way that, only terms with degree less than n remain. To do so, we pick, $m(x)$ as the negative of the quotient when x^K is divided by $f(x)$. Which eventually leaves us with $x^K \bmod f(x)$

Hence the final solution

Find $x^K \bmod f(x)$ and then convert it to recurrence, which will have only terms R_0, R_1, \dots, R_{n-1}

The way to do this is:

- Figure out how to calculate poly mod poly. (remainder when a polynomial is divided by another polynomial)
- Use binary exponentiation to find $x^K \bmod f(x)$

It is easy to see that there will be $O(\log K)$ steps, with each step involving poly mod poly.

Hence using FFT, this can be done in $O(N \log N \log K)$

"Wait! You didn't tell how to do Poly mod Poly!"

Now, I won't be explaining how to do Poly mod Poly, because it is pretty easy to do naively in $O(N^2)$, and very complex using FFT in $O(N \log N)$ — [You can read about it here](#).

However, you can just use Poly mod Poly as a **black box** and use this [code library](#) by **adamant** (Although it is not very optimized, but it is good for basic use)


Practice Problems

[RNG Codechef](#)

You can also try the problems from the [matrix exponentiation mashup](#) using this technique, but it'll be an overkill.

▲ +299 ▼ ☆

 [demoralizer](#)

 17 months ago

 [10](#)



Comments (10)

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vishaaaal

17 months ago, <#> | ☆

← Rev. 2

▲ +43 ▼

Thanks for this tutorial!

Also, another great [tutorial](#), which I used to learn about Berlekamp-Massey.

A [problem](#) which appeared in an Edu Round.
Also one more [practice problem](#) and [editorial](#).

→ [Reply](#)



Golovanov399 

17 months ago, <#> | ☆

← Rev. 2

▲ +48 ▼

Given the characteristic polynomial, one can calculate the K -th term faster than you suggested, that is, without poly mod poly (but with the same time complexity). See [my comment](#) from the other blog.

Also, [here](#) one can check their implementation

→ [Reply](#)

Hmm, I was wondering if there is any meaningful way to explain the $x^k \bmod f(x)$ part with as little direct manipulations with coefficients as possible. I know that we may just directly show it by comparing rows of C^{n+1} and C^n matrices, but it's somewhat ugly to me. And the approach in this blog doesn't seem rigorous enough. What if we look on it from generating functions perspective?

For $g(x) = \sum_{i=0}^{\infty} R_i x^i$, the linear recurrence essentially means that

$$g(x)t(x) = g(x)t(x) \bmod x^n,$$

where $t(x) = 1 - \sum_{i=1}^n C_i x^i$. The modulo part here means that $g(x)t(x)$ has zero coefficient (\iff the recurrence stands) near x^k for every $k \geq n$ and for $k < n$ the coefficient is not restricted in any way. Thus, $g(x)$ may be represented as

$$g(x) = \frac{a(x)}{t(x)}$$

where $a(x) = g(x)t(x) \bmod x^n$. Since the product is taken modulo x^n , only first n coefficients of $g(x)$ matter for the definition of $a(x)$. Note that $t(x) = x^n f(x^{-1})$ where $f(x)$ is the characteristic function defined in this blog. Because of that, $d(x) = g(x^{-1})$ might be represented as

$$d(x) = \frac{a(x^{-1})}{t(x^{-1})} = \frac{x^n a(x^{-1})}{f(x)}$$

Noteworthy, $x^k d(x)$ has R_k as the coefficient near x^0 . On the other hand, $x^t f(x) d(x) = x^{n+t} a(x^{-1})$ always has zero coefficient near x^0 for $k \geq 0$, as $a(x)$ is the polynomial of degree at most $n - 1$.

Now, to find the linear combination of R_0, \dots, R_{n-1} that yields R_k is the same as to find a polynomial $p(x)$ of degree at most n such that coefficients near x^0 in $x^k d(x)$ and $p(x) d(x)$ are the same.

But since for $x^t f(x) d(x)$ this coefficient is 0, we may conclude that polynomials $p(x)$ and $p(x) \bmod f(x)$ yield the same coefficient near x^0 and, thus, it is indeed safe to go with $p(x) = x^k \bmod f(x)$.

→ [Reply](#)



adamant

17 months ago, # ^ | ☆

▲ +23 ▼



demoralizer

Thanks for this Mathematical Proof of the technique!

Before this, I always thought of this technique as a fast way of substituting the value of the highest order term repeatedly, and assumed that the operations involved are just "coincidentally" the same as poly mod poly.

→ [Reply](#)

16 months ago, # ^ | ☆

← Rev. 3

▲ +5 ▼

This digression seems rather unnecessary to me, the idea given in TLE's blog works just fine.

For the recurrence $a_k = \sum_{i=1}^n c_i a_{k-i}$. Define the linear functional $T : \mathbb{F}[x] \rightarrow \mathbb{F}$ as $T(x^k) = a_k$.



islingr

Now, if f is the characteristic polynomial defined in the blog, $T(x^r f) = 0$ for any $r \geq 0$ by definition of the recurrence. That is, $\text{span}_{r \geq 0}(x^r f) \subseteq \ker T$. This span is nothing but the polynomial multiples of f , call it $f\mathbb{F}[X]$.

You're done here, you see that $T(p) = T(q)$ for any $p - q \in f\mathbb{F}[x]$ so $T(x^k) = T(x^k \bmod f)$.

If you wish, you can view it as a linear functional over $\mathbb{F}[x]/f\mathbb{F}[x]$ but that's just adding more needless jargon. Really all we did here was write the "direct manipulations" here in a linear algebra setting.

→ [Reply](#)

16 months ago, # ^ | ☆

← Rev. 2

▲ 0 ▼



adamant

Yeah, that's nice way to see it. As I understand, $T(p) = [x^0]p(x)g(x^{-1})$ in my terms... So, I essentially do the same stuff, but a bit overly verbose, perhaps.

→ [Reply](#)

17 months ago, # | ☆

← Rev. 6

▲ +19 ▼



oToToT

The first problem I tried with this technique is [this atcoder problem](#) by **E869120**. (The whole series of problems are good for beginners.)

The problems goes like:

Given N, K , find the number (mod 998244353) of possible non-negative sequence A_i with size N satisfying $\forall 1 \leq l \leq r \leq N$,

$\text{sum}(a_l, a_{l+1}, \dots, a_r) \wedge (l - r \geq 1) \geq 1$.

→ [Reply](#)

17 months ago, <#> | ☆

▲ -48 ▼



MarkZuckerberg

demoralizer finally did it.

I WAS THE FIRST PERSON WHO REQUESTED IT !

<https://codeforces.com/blog/entry/83164?#comment-704678>

→ [Reply](#)



dthompson

17 months ago, <#> | ☆

▲ -37 ▼

Are you sure Berlekamp Massey isn't better? There are a lot of log factors in your code too.

→ [Reply](#)

17 months ago, <#> ^ | ☆

▲ +16 ▼



nor

- Berlekamp Massey has a different purpose: "interpolate" a linear recurrence from the first few values, while the blog concerns itself with just computing some element of a given linear recurrence.
- Why is having log factors bad? N^{100} is worse than $N \log^{100} N$, isn't it?

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