# Parallelized Polynomial Multiplication (MultPoly)

Yaxin Chen (yc3995)

November 17, 2021

## Introduction

In this project, I will parallelize two algorithms for multiplying two polynomials and compare their runtime. One is the native approach with a time complexity of  $O(n^2)$ , where n is the degree of the polynomial; the other utilizes fast fourier transform (FFT) and has a time complexity of  $O(n \log n)$ .

#### **Brute-Force Polynomial Multiplication**

A degree-(n-1) polynomial can be represented by an n-element array storing its coefficients. Suppose we have array A representing polynomial  $a(x) = \sum_{i=0}^{n-1} A[i]x^i$  and B representing polynomial  $b(x) = \sum_{i=0}^{n-1} B[i]x^i$ , the array C for the product of a(x) and b(x) can be calculated by

```
for i \leftarrow 0 to n-1 do
for j \leftarrow 0 to n-1 do
C[i + j] \leftarrow C[i + j] + A[i] * B[j];
end for
end for
```

**Parallelization:** To parallel the above calculation, I plan to use MapReduce framework. The mapper takes pairs of coefficient of two input polynomials (A[i], B[j]), multiplies them, and sends (key: i+j, value:  $A[i]^*B[j]$ ) to the reducer. The reducer sums the received product and gives output coefficient at index (i+j).

#### Polynomial Multiplication via FFT

The polynomial multiplication can be speed up to  $O(n \log n)$  by fast fourier transforming the input polynomials, multiplying them and the inverse fourier transforming the product.

The discrete fourier transform (DFT) of an n-element sequence A is another n-element sequence P given by

$$P[m] = \sum_{k=0}^{n-1} A[k]\omega_n^{mk}, \ m = 0, 1, ..., n-1$$

where  $\omega_n = e^{2\pi i/n}$  is the primitive  $n^{th}$  root of unity.

For  $0 \le m < n/2$ , DFT satisfies

$$P[m] = P_1[m] + \omega^m P_2[m] \tag{1}$$

$$P[n/2 + m] = P_1[m] - \omega^m P_2[m]$$
(2)

where

$$P_1[m] = \sum_{k=0}^{n/2-1} A[2k]\omega_n^{2mk}$$
$$P_2[m] = \sum_{k=0}^{n/2-1} A[2k+1]\omega_n^{2mk}$$

FFT utilizes the above property (Eq 1, 2). The algorithm for FFT is shown in Algorithm 1.

Algorithm 1 Fast Fourier Transform

```
1: procedure FFT(A, n, \omega)
         if n = 1 then return A;
 2:
         else
 3:
             for \mathbf{k} \leftarrow 0 to \mathbf{n}/2 - 1 do
 4:
 5:
                  A_1[k] = A[2k]
 6:
                  A_2[k] = A[2k+1]
             end for
 7:
             P_1 \leftarrow FFT(A_1, n/2, \omega^2)
 8:
             P_2 \leftarrow FFT(A_2, n/2, \omega^2)
 9:
             for m \leftarrow 0 to n - 1 do
10:
                  P[m] \leftarrow P_1[m \mod (n/2)] + \omega^m P_2[m \mod (n/2)]
11:
             end for
12:
         end if
13:
14: end procedure
```

**Parallelization:** Since the calculation of  $P_1$  and  $P_2$  are independent with each other, their calculation can be parallelized. This part can go parallel to a certain depth and I will investigate the speedup versus the depth. What's more, binary-exchange algorithm and transpose algorithm [1] can parallelize FFT with a better granularity, and I will implement them in Haskell and do some comparisons.

### References

- 1. https://courses.engr.illinois.edu/cs554/fa2015/notes/13\_fft\_8up.pdf
- 2. http://www.cs.toronto.edu/~denisp/csc373/docs/tutorial3-adv-writeup.pdf
- 3. https://cse.hkust.edu.hk/mjg\_lib/Classes/COMP3711H\_Fall14/lectures/DandC\_Multiplication\_ Handout.pdf