Talker: 張志謙

PQC Final Report

2022/6/17

Definition:

要計算兩個多項式A(x), B(x),分別將其分成k部分(每部分長度固定),再對這些部分執行運算。隨著k的增長,可以組合許多乘法子運算,從而降低算法的整體複雜度,然後再次使用Toom-k算法遞歸計算乘法子運算,依此類推。

流程:

- 1. 拆分&求值⇒ Linear Transformation
- 2. 點乘
- 3. 插值⇒ Linear Transformation

拆分&求值:

Let
$$A(x) = a_{n-1}x^{n-1} + \dots, a_0, B(x) = b_{n-1}x^{n-1} + \dots, b_0$$
, and we set $y = x^{\frac{n}{k}}$, $\alpha_{k-1} = a_{n-1}x^{\frac{n}{k}-1} + \dots + a_{n-\frac{n}{k}}, \alpha_{k-2} = a_{n-1-\frac{n}{k}}x^{\frac{n}{k}-1} + \dots + a_{n-\frac{2n}{k}}, \dots$
$$\beta_{k-1} = b_{n-1}x^{\frac{n}{k}-1} + \dots + b_{n-\frac{n}{k}}, \dots$$
 then
$$\begin{cases} A(y) = \alpha_{k-1}x^{k-1} + \dots + \beta_0 \\ B(y) = \beta_{k-1}x^{k-1} + \dots + \beta_0 \end{cases}$$

拆分&求值:

choose 2k-1 points $p_0,...,p_{2k-2}$ on A(y)

$$\begin{bmatrix} A(p_0) \\ A(p_1) \\ \vdots \\ A(p_{2k-2}) \end{bmatrix} = \begin{bmatrix} (p_0)^0 & (p_0)^1 & \dots & (p_0)^{k-1} \\ (p_1)^0 & (p_1)^1 & \dots & (p_1)^{k-1} \\ \vdots \\ (p_{2k-2})^0 & (p_{2k-2})^1 & \dots & (p_{2k-2})^{k-1} \end{bmatrix} \begin{bmatrix} \alpha_0 \\ \alpha_1 \\ \vdots \\ \alpha_{k-1} \end{bmatrix}$$

插值:

Use
$$C(p_0), ..., C(p_{2k-2})$$
 to recover $C(y) = c_{2k-2}y^{2k-2} + , ..., c_0$

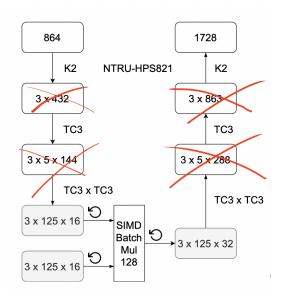
$$\begin{bmatrix}
C(p_0) \\
C(p_1) \\
... \\
... \\
C(p_{2k-2})
\end{bmatrix} = \begin{bmatrix}
(p_0)^0 & (p_0)^1 & ... & (p_0)^{k-1} \\
(p_1)^0 & (p_1)^1 & ... & (p_1)^{k-1} \\
... & ... & ... & ... \\
(p_{2k-2})^0 & (p_{2k-2})^1 & ... & (p_{2k-2})^{k-1}
\end{bmatrix} \begin{bmatrix}
c_0 \\
c_1 \\
... \\
... \\
c_{2k-2}
\end{bmatrix}$$

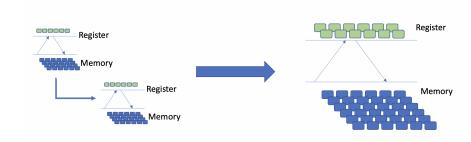
插值:

Use
$$C(p_0),...,C(p_{2k-2})$$
 to recover $C(y)=c_{2k-2}y^{2k-2}+,...,c_0$

$$\begin{bmatrix}c_0\\c_1\\.\\.\\c_{2k-2}\end{bmatrix}=$$

$$\begin{bmatrix} (p_0)^0 & (p_0)^1 & \dots & (p_0)^{k-1} \\ (p_1)^0 & (p_1)^1 & \dots & (p_1)^{k-1} \\ \dots & \dots & \dots & \dots \\ (p_{2k-2})^0 & (p_{2k-2})^1 & \dots & (p_{2k-2})^{k-1} \end{bmatrix}^{-1} \begin{bmatrix} C(p_0) \\ C(p_1) \\ \vdots \\ C(p_{2k-2}) \end{bmatrix}$$





Consider all column vectors in
$$\begin{bmatrix} (p_0)^0 & (p_0)^1 & \dots & (p_0)^{k-1} \\ (p_1)^0 & (p_1)^1 & \dots & (p_1)^{k-1} \\ \dots & \dots & \dots & \dots \\ (p_{2k-2})^0 & (p_{2k-2})^1 & \dots & (p_{2k-2})^{k-1} \end{bmatrix} :$$

if $\forall i, j$ satisfy $p_i \neq p_j$, then these vectors are independent. Hence, Toom-Cook multiplication and interpolation are fundamentally linear maps of the form:

$$\begin{cases}
\mathsf{TC}_k : R_{\frac{n}{k}-1}^k(x) \to R_{\frac{n}{k}-1}^{2k-1}(x) \\
\mathsf{TC}_k^{-1} : R_{\frac{2n}{k}-2}^{2k-1}(x) \to R_{\frac{2n}{k}-2}^{2k-1}(x)
\end{cases}$$

$$\hat{TC}(A(x)) = TC_{k_{\eta}}(TC_{k_{\eta-1}}(...TC_{k_{1}}(A(x))))$$

$$C(x) = TC_{k_1}^{-1}(TC_{k_2}^{-1}(...TC_{k_n}^{-1}(\theta)))$$

Definition:

$$O_{n} = \begin{bmatrix} 0 & \dots & 0 \\ \cdot & \dots & \cdot \\ \cdot & \dots & \cdot \\ 0 & \dots & 0 \end{bmatrix}_{n \times n}, I_{n} = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \cdot & \dots & \dots & \cdot \\ \cdot & \dots & \dots & \cdot \\ 0 & \dots & \dots & 1 \end{bmatrix}_{n \times n},$$

$$T_{n} = \begin{bmatrix} I_{n} & & & & \\ I_{n} & I_{n} & & & \\ & & I_{n} & & \\ & & & & & \end{bmatrix}_{n \times n}$$

inprovement
$$T_{\frac{n}{2}} = \begin{bmatrix} I_{\frac{n}{2}} & I_{\frac{n}{2}} \\ I_{\frac{n}{2}} & I_{\frac{n}{2}} \end{bmatrix}_{\frac{3}{2}n \times n}$$

$$\Rightarrow \begin{bmatrix} T_{\frac{n}{4}} & & & \\ & T_{\frac{n}{4}} & & \\ & & & \end{bmatrix}_{\frac{n}{4}}$$

$$\begin{bmatrix} I_{\frac{n}{4}} & I_{\frac{n}{4}} \\ I_{\frac{n}{4}} & I_{\frac{n}{4}} \end{bmatrix}$$

$$\frac{3}{2}n\times n$$

Reference

Time-memory trade-off in Toom-Cook multiplication: an application to module-lattice based cryptography