

August 6th 2024

Introduction to Additive NTT

Recall that NTT over a large finite field \mathbb{F}_p works by transforming monomial basis coeffs of a univariate polynomial to Lagrange Basis coefficients over a carefully selected domain (read: evals of said polynomial over some input domain D)

- Suppose the transform is for univariate polynomials of degree $< 2^l$, i.e. $\mathbb{F}_p[X]^{< 2^l}$
- let $f(x) \in \mathbb{F}_p[X]^{< 2^l} = a_0 + a_1 X + \dots + a_{2^l-1} X^{2^l-1}$
- NTT transforms $\vec{a} = (a_i)_{i=0, \dots, 2^l-1}$ into $\vec{b} = (b_i)_{i=0, \dots, 2^l-1}$
where $b_i = f(d_i)$ and $D = \{d_i \mid i \in \{0, \dots, 2^l-1\}\} \subseteq \mathbb{F}_p$
is the carefully selected domain
- When $D = \langle \omega_{2^l} \rangle$, i.e. the (multiplicative) subgroup of \mathbb{F}_p generated by the 2^l -th root of unity, traditional NTT and iNTT admit, respectively, $\Theta(l \cdot 2^l)$ algorithms to perform change of basis from \vec{a} to \vec{b} and \vec{b} to \vec{a} .
- For binary fields, however, this 2^l -th root of unity will not exist, and we would like to find a new way to perform a change of basis in time $\Theta(l \cdot 2^l)$

Enter Additive NTT

Outline

- ① Subspace Polynomials
- ② Novel Polynomial Basis
- ③ Recursive Structure
- ④ Algorithmic Structure
- ⑤ The forward transform
- ⑥ Linear Codes
- ⑦ Reed Solomon Encoding
- ⑧ Quick Note on inverse transform and RS Decoding

① Subspace Polynomials

- Fix a binary field L of degree r over \mathbb{F}_2 (i.e. $L \cong \mathbb{F}_{2^r}$)
- Fix $B = (\beta_0, \dots, \beta_{r-1})$ as an r -dimensional \mathbb{F}_2 -basis for L ($\beta_i \in L$, β_i 's are \mathbb{F}_2 -linearly independent)
- Let $U_i = \langle \beta_0, \dots, \beta_{i-1} \rangle$ be the subspace of L generated by the first i basis vectors for $i=0, \dots, r$
- Defn: A subspace polynomial $w(x) \in L[x]$ splits completely over L and its roots, each of multiplicity 1, form an \mathbb{F}_2 -linear subspace of L
- Define $w_i(x) := \prod_{u \in U_i} X+u$ and $\hat{w}_i(x) := \frac{w_i(x)}{w_i(\beta_i)}$
C normalized variant. $\hat{w}_i(\beta_i) = 1$
- Note: $w_0(x) := \prod_{u \in U_0} X+u = X+0 = X$
- Note: $\deg(w_i(x)) = 2^i$
- $w_{i+1}(x) = w_i(x) \cdot w_i(x + \beta_i)$ by lemma
- $$\prod_{u \in U_{i+1}} X+u = \left[\prod_{u \in U_i} X+u \right] \cdot \left[\prod_{u \in U_i} X + (\beta_i + u) \right]$$

$$= w_i(x) \cdot [w_i(x) + w_i(\beta_i)]$$
by lemma

- For $b \in \{0, \dots, 2^r - 1\}$, let b_0, \dots, b_{r-1} denote its bits such that $b = \sum_{i=0}^{r-1} b_i \cdot 2^i$

Then we denote as $w_b = [B_0, B_1, \dots, B_{r-1}] \cdot [b_0, b_1, \dots, b_{r-1}]$
 $\quad \quad \quad // \text{basis elements} \quad \quad \quad \mathbb{F}_2 - \text{vector}$

Thus $L = \{w_b \mid b \in \{0, \dots, 2^r - 1\}\}$

- Note: $\beta_i = w_{z_i}$

Summary of important results related to subspace polynomial

- $\forall x, y \in L \text{ , } \forall k \in \{0, \dots, r-1\}$

$$W_k(x+y) = W_k(x) + W_k(y)$$

$$\hat{W}_K(x+y) = \hat{W}_K(x) + \hat{W}_K(y)$$

- $\forall k \in \{0, \dots, r-1\}, \forall b \in \{0, \dots, 2^k - 1\}$

$$W_k(w_b) = 0 \quad | \quad w_b \in U_k$$

$$W_k(w_b) = 0 \quad \Rightarrow W_x(w_b)$$

- ② $\forall k \in \{0, \dots, r-2\}$

$$W_{K+1}(x) = W_K(x) \cdot W_K(x + \beta_i)$$

②

Novel Polynomial Basis

- We have fixed L as a degree r extension of \mathbb{F}_2 with basis $B = (B_0, \dots, B_{r-1})$
 - Now, fix $l \in \{0, \dots, r-1\}$
 - We denote as $L[x]^{<2^l}$ the ring of polynomials with L -coefficients of degree $< 2^l$.
- $$L[x]^{<2^l} = \left\{ P(x) \in L[x] \mid \deg(P) < 2^l \right\}$$
- $L[x]^{<2^l}$ is a 2^l -dimensional vector space over L , and therefore has a basis of size 2^l .
 - A natural basis would be the (univariate) monomial basis $(1, x, x^2, \dots, x^{2^l-1})$. However, we do not know a $O(l \cdot 2^l)$ algorithm to change basis b/w monomial and a Lagrange basis (i.e. we do not have a $O(l \cdot 2^l)$ algo that gives us 2^l unique evaluations given monomial coeffs as input)
 - Define $X_i(x) = \prod_{k=0}^{l-1} (W_k(x))^{i_k}$ where $i \in \{0, 1\}^l$
 - Note $\deg(X_i(x)) = i$
 - Now $(X_0(x), \dots, X_{2^l-1}(x))$ admits an L -basis for $L[x]^{<2^l}$ called the NOVEL POLYNOMIAL BASIS

③ Recursive Structure of Nörl Polynoial Basis

- We have fixed binary field L of degree r over \mathbb{F}_2 with basis $\beta = (\beta_0, \dots, \beta_{r-1})$
- We have fixed $l \in \{0, \dots, r-1\}$
- Fix $D(x) = d_0 X_0(x) + \dots + d_{2^l-1} X_{2^l-1}(x) \in L[x]^{2^l}$
- Define $\Delta_i^m(x) = d_m \quad \forall m \in \{0, \dots, 2^l-1\}$
- $$\Delta_i^m(x) = \Delta_{i+1}^m(x) + \hat{W}_i(x) \cdot \Delta_{i+1}^{m+2^i}(x)$$

$$\forall i \in \{0, \dots, l-1\} \quad \forall m \in \{0, \dots, 2^i-1\}$$
- $\deg_x(\Delta_i^m(x)) \leq 2^{l-i}-1 \quad \forall m \in \{0, \dots, 2^i-1\}$
- Claim without proof: $\Delta_0^0(x) = D(x)$

Ex $l=3$, $D(x) = \sum_{i=0}^7 d_i X_i(x)$

$$= [d_0, \dots, d_7] \cdot [X_0(x), \dots, X_7(x)]$$

$$= [d_0 + d_4 \hat{W}_2(x) + \hat{W}_1(x) [d_2 + d_6 \hat{W}_2(x)]]$$

$$+ \hat{W}_0(x) [d_1 + d_5 \hat{W}_2(x) + \hat{W}_1(x) [d_3 + d_7 \hat{W}_2(x)]]$$

$$= [\Delta_2^0(x) + \hat{W}_1(x) \Delta_2^2(x)]$$

$$+ \hat{W}_0(x) [\Delta_2^1(x) + \hat{W}_1(x) \Delta_2^3(x)]$$

$$= \Delta_0^0(x) + \hat{W}_0(x) \Delta_1^1(x) = \Delta_0^0(x)$$
□

• Lemma: $\Delta_i^m(x+y) = \Delta_i^m(x) \quad \forall y \in \{\omega_b\}_{b=0}^{2^i-1}$

• Pf By induction on i

$$\boxed{\text{Bc}} \quad i = l-1 \quad \Delta_{l-1}^m(x) \\ = \Delta_l^m(x) + \hat{W}_{l-1}(x) \Delta_{l-1}^{m+2^{l-1}}(x) \\ = d_m + \hat{W}_{l-1}(x) d_{m+2^{l-1}}$$

similarly $\Delta_{l-1}^m(x+y) = d_m + \hat{W}_{l-1}(x+y) d_{m+2^{l-1}}$

but $\hat{W}_{l-1}(x+y) = \hat{W}_{l-1}(x)$ when $y \in \{\omega_b\}_{b=0}^{2^{l-1}-1}$

If Assume true for $i = l+1$

$$\boxed{\text{Is}} \quad \Delta_c^m(x+y) \\ = \Delta_{c+1}^m(x+y) + \hat{W}_c(x+y) \Delta_{c+1}^{m+2^c}(x+y) \\ = \Delta_{c+1}^m(x) + \hat{W}_c(x) \Delta_{c+1}^{m+2^c}(x) \quad \text{by If and } \hat{W}(x+y) = \hat{W}(x)$$

3,5

Algorithmic Structure

- We have fixed binary field L of degree r over \mathbb{F}_2 with basis $\vec{\beta} = (\beta_0, \dots, \beta_{r-1})$
- We have fixed a parameter $l \in \{0, \dots, r-1\}$ and consider the 2^l -dimensional L -vector space $L[X]^{< 2^l}$ (univariate polys over L with degree $< 2^l$)
- We have fixed $D(x) = \sum_{i=0}^{2^l-1} d_i X_i(x) \in L[X]^{< 2^l}$

Our objective now is to exploit the recursive structure of the novel polynomial basis to build an algorithmic structure for our goal of efficient change-of-basis

- Fix a "Shift parameter" $s \in \{0, \dots, 2^l-1\}$ //more on this later, if easier imagine $s=0$

Defn

$$\forall i \in \{0, \dots, l-1\} \quad \forall m \in \{0, \dots, 2^i-1\}$$

$$\Psi(i, m, s) := \left\{ \Delta_i^m(w_c + w_s) \mid c \in \left\{ b \cdot 2^i \right\}_{b=0}^{2^{l-i}-1} \right\}$$

is a set of size 2^{l-i}

Defn

$$\Psi(l, m, s) := \{d_m\} \quad \text{is a singleton set}$$

size = $2^{l-1} = 2^0 = 1$

Our goal is to transform $\Psi(l, 0, s), \dots, \Psi(l, 2^l-1, s)$ to $\Psi(0, 0, s)$

and vice-versa

$$\begin{array}{c} \downarrow \\ \{d_0\} \end{array} \quad \begin{array}{c} \downarrow \\ \{d_{2^l-1}\} \end{array}$$

$$\left\{ \Delta_0^0(w_0), \Delta_0^0(w_1), \dots, \Delta_0^0(w_{2^l-1}) \right\}$$

When $s=0$

- Fix $i \in \{0, \dots, l-1\}$

We will now show how you can relate

$$\forall m \in \{0, \dots, 2^i-1\} \quad \Psi(i, m, s) \leftrightarrow \forall n \in \{0, \dots, 2^{i+1}-1\} \quad \Psi(i+1, n, s)$$

• Fix $m \in \{0, \dots, 2^i - 1\}$

$$\Psi(i, m, s) = \Psi^0(i, m, s) \cup \Psi^1(i, m, s)$$

where

$$\Psi^0(i, m, s) = \left\{ \Delta_i^m(w_c + w_s) \mid c \in \{b \cdot 2^i\}_{b=0}^{2^{i-1}-1} \right\}$$

$$\Psi^1(i, m, s) = \left\{ \Delta_i^m(w_c + w_{2^i} + w_s) \mid c \in \{b \cdot 2^i\}_{b=0}^{2^{i-1}-1} \right\}$$

• Fix $b \in \{0, \dots, 2^{i-1}-1\}$ let $c = b \cdot 2^i$

$$\begin{aligned} \Psi^0(i, m, s)[b] &= \Delta_i^m(w_c + w_s) = \underbrace{\Delta_{i+1}^m(w_c + w_s)}_{= \Psi(i+1, m, s)[b]} + \hat{W}_i(w_c + w_s) \underbrace{\Delta_{i+1}^{m+2^i}(w_c + w_s)}_{= \Psi(i+1, m+2^i, s)[b]} \\ &= \Psi(i+1, m+2^i, s)[b] \end{aligned}$$

$$\begin{aligned} \Psi^1(i, m, s)[b] &= \Delta_i^m(w_c + w_{2^i} + w_s) = \Delta_{i+1}^m(w_c + w_{2^i} + w_s) + \hat{W}_i(w_c + w_{2^i} + w_s) \underbrace{\Delta_{i+1}^{m+2^i}(w_c + w_{2^i} + w_s)}_{= \Psi(i+1, m+2^i, s)[b]} \\ &= \Delta_{i+1}^m(w_c + w_s) + \hat{W}_i(w_c + w_{2^i} + w_s) \underbrace{\Delta_{i+1}^{m+2^i}(w_c + w_s)}_{= \Delta_{i+1}^m(w_c + w_s) + (\hat{W}_i(w_c + w_s) + \hat{W}_i(\beta_i)) \Delta_{i+1}^{m+2^i}(w_c + w_s)} \\ &= \Delta_{i+1}^m(w_c + w_s) + (\hat{W}_i(w_c + w_s) + 1) \Delta_{i+1}^{m+2^i}(w_c + w_s) \\ &= \Delta_i^m(w_c + w_s) + \underbrace{\Delta_{i+1}^{m+2^i}(w_c + w_s)}_{= \Psi(i+1, m+2^i, s)[b]} \\ &= \Psi(i+1, m+2^i, s)[b] \end{aligned}$$

• Summary: Fix $i \in \{0, \dots, k-1\}$

For each $m \in \{0, \dots, 2^i - 1\}$ and $b \in \{0, \dots, 2^{i-1}\}$

$(\Psi[i, m, s][b], \Psi[i, m, s][b + 2^{i-1}])$ can be derived from $(\Psi[i+1, m, s][b], \Psi[i+1, m+2^i, s][b])$

$x_0 \quad x_1 \quad (\text{also vice versa})$

$y_0 \quad y_1$

The process to convert from (y_0, y_1) into (x_0, x_1)

Structurally looks like

$$x_0 = y_0 + \boxed{\quad} \cdot y_1$$

$$x_1 = x_0 + y_1$$

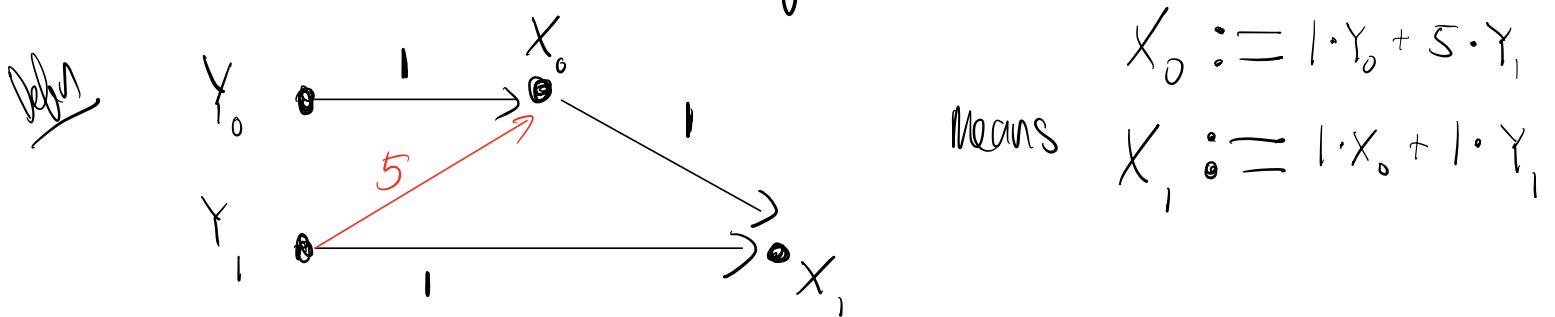
where $\boxed{\quad}$ is some factor depending on
 i, b that's called in our codebase the "twiddle" factor

$$\boxed{\quad}_{i,b} = \hat{w}_i (w_s + w_{b \cdot 2^i})$$

④ The forward transform

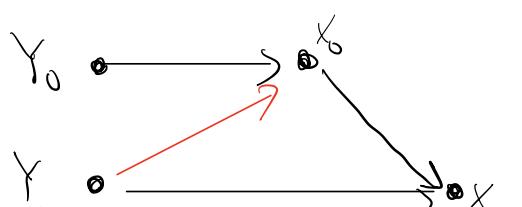
- We have fixed
 - binary field L of degree r over \mathbb{F}_2 with basis $B = (B_0, \dots, B_{r-1})$
 - Parameter $\ell \in \{0, \dots, r-1\}$ defining our polynomial ring $L[x]^{2^{\ell}}$
 - $D(x) = \sum_{k=0}^{2^\ell - 1} d_k \cdot X_k(x) \in L[x]^{2^\ell}$
- Our goal is to take as input $d = [d_0, \dots, d_{2^\ell - 1}]$, $s \in \{0, \dots, 2^\ell - 1\}$
 and output $[D(w_0 + w_s), \dots, D(w_{2^\ell - 1} + w_s)]$
 in time $O(\ell \cdot 2^\ell)$

Let us now introduce the "butterfly" diagram



Notes: This is a DAG, so one can top-sort and assign nodes values in that order

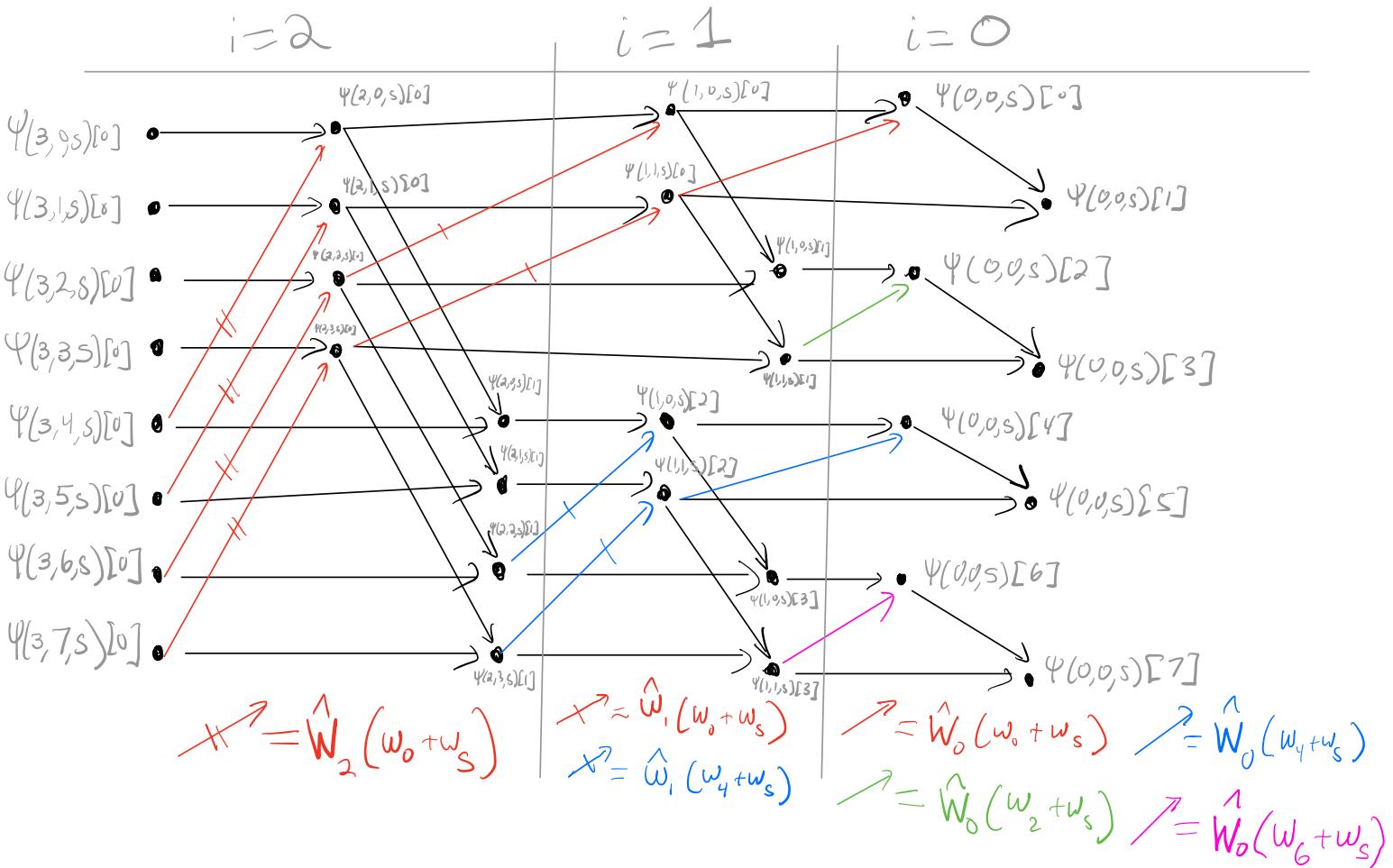
- A node's value is the sum of its edge-weighted in-node values (weight is multiplicative factor)
- If the edge-weight is 1, we omit explicitly writing this
- Sometimes, we use colors to indicate edge weights



is the same diagram

where $\textcolor{red}{\rightarrow} = 5$

~~Ex~~ Forward transform when $\ell=3$, arbitrary S



The diagram above suffices as a complete description of the algorithm

Let's talk about how to compute the "fudge-factors" i.e. $\hat{W}_i(w_{j * \text{block-size}} + w_s)$

- For simplicity, let's assume $S = 2^\ell$, so, $w_s = w_{2^\ell} = \beta_\ell$

- Precompute constants $W_0(\beta_0), \dots, W_0(\beta_{\ell-1}), W_0(\beta_\ell)$

- For $i = 1, \dots, \ell - 1$:

- Compute $W_i(\beta_i), \dots, W_i(\beta_\ell)$

using $W_i(\beta_j) := W_{i-1}(\beta_{j-1}) \cdot [W_{i-1}(\beta_{j-1}) + W_{i-1}(\beta_{i-1})]$

- For each $i \in \{0, \dots, \ell - 1\}$, For each $j \in \{i, \dots, \ell\}$, $\hat{W}_i(\beta_j) = W_i(\beta_j) / W_i(\beta_i)$

Given
 $\hat{W}_i(\beta_j)$ $\forall i \in \{0, \dots, \ell - 1\}$
 $\forall j \in \{i, \dots, \ell\}$

We can use

- $\hat{W}_i(x+y) = \hat{W}_i(x) + \hat{W}_i(y)$

to calculate $\hat{W}_i(w_k)$

$\forall i \in \{0, \dots, \ell - 1\}$

$\forall k \in \{0, \dots, 2^\ell - 1\}$

⑤ Linear Codes

Defn: A linear code is a code that is a subspace of a vector space over a finite field \mathbb{F}

Namely if our code is $C \subseteq \mathbb{F}^n$, and $x, y \in C$, then

$$\forall a, b \in \mathbb{F}, \quad ax + by \in C$$

(setting $a=1, b=-1$)

$$x - y \in C$$

Defn: The Hamming Distance function $\Delta: \mathbb{F}^n \times \mathbb{F}^n \rightarrow \{0, \dots, n\}$

$(x, y) \mapsto$ The number of indices $i \in \{0, \dots, n\}$ s.t.

$$x_i \neq y_i$$

Defn: The dimension of code C is simply the dimension of C (of the linear subspace)

It is usually denoted by k

Recall if C is a k -dimensional subspace of \mathbb{F}^n , then \exists matrix $G \in \mathbb{F}^{k \times n}$

$$\begin{array}{c} \text{msg} \\ \hline \text{length } k \end{array} \cdot \underbrace{\begin{array}{c} G \\ \hline n \end{array}}_{\text{length } n} = \underbrace{\text{codeword}}_{n}$$

- C is the span of G_0, \dots, G_{k-1} , the k rows of G
- a codeword is a linear combination of the rows of G where the k coeffs is the message

Technical note: If you're curious, look up "parity check matrix" H . It's another way to define a linear code.

Defn: Distance of a linear code is $\min_{x, y \in C} \{ \Delta(x, y) \}$

or equivalently $\min_{x \in C} \{ \Delta(x, 0) \}$

Challenge: convince yourself of this equivalence.

Note: a linear code $C \subseteq \mathbb{F}_q^n$ of dimension K , distance d , is denoted as a $[n, K, d]_q$ code

Lemma: Singleton Bound states that every $[n, K, d]_q$ linear code satisfies the inequality

$$d \leq n+1-K$$

when a code achieves equality it's called MDS

Defn Maximum Distance Separable (MDS) is when $d = n+1-K$

Defn The rate of a code $R := \frac{K}{n} \in (0, 1]$

$R \uparrow$ redundancy \downarrow space efficiency \uparrow

$R \downarrow$ redundancy \uparrow space efficiency \downarrow

Sometimes we care about inv-rate $\in [1, +\infty)$

$$\frac{n}{R}$$

⑥ Reed Solomon Codes:

- Fix some domain $D = \{x_0, \dots, x_{n-1}\} \subseteq \mathbb{F}_q$
- We can define the RS code $[n, K, n+1-K]_q$ like so

$$C = \left\{ \begin{bmatrix} P(x_0), \dots, P(x_{n-1}) \end{bmatrix} \mid P(x) \in \mathbb{F}_q[x]^{< K} \right\}$$

~~Facts~~: RS codes are linear.

why? $\begin{bmatrix} P(x_0) + Q(x_0), \dots, P(x_{n-1}) + Q(x_{n-1}) \end{bmatrix}$
 $= \begin{bmatrix} (P+Q)(x_0), \dots, (P+Q)(x_{n-1}) \end{bmatrix}$

Step. If this doesn't sit right with you, write out $P(X) = a_0 + a_1 X + a_2 X^2$
 $Q(X) = b_0 + b_1 X + b_2 X^2$
 $R(X) = (a_0+b_0) + (a_1+b_1)X + (a_2+b_2)X^2$
and convince yourself $\forall z, P(z)+Q(z)=R(z)$

- RS codes are MDS, so $d = n+1-K$ always

Why? polynomial magic. Two distinct polynomials of deg $\leq K-1$ can agree on at most $K-1$ places. They disagree at $\geq n-(K-1) = n+1-K$ places.

No reason to write $[n, K, d]_q$, we instead write $RS(n, K)_q$

What happens when $n=k$ (and $d=1$)?

No, this does not mean that msg is its own codeword

$$\begin{array}{ccc} m & \xrightarrow{\quad} & C \\ m \in \mathbb{F}_q^K & \xrightarrow{\text{encoding}} & c \in C \subseteq \mathbb{F}_q^n \end{array}$$

univariate!

- We are given the msg as some polynomial P
coeffs FOR SOME BASIS

let's say basis is $(1, x, x^2, \dots, x^{K-1})$

- The output is "K evaluations of P over domain D "
or... coeffs for P over the "univariate Lagrange basis w.r.t. D "

so basis is $(L_{x_0}(x), L_{x_1}(x), \dots, L_{x_{K-1}}(x))$

where $L_{x_i}(x) = \begin{cases} 1, & x = x_i \\ 0, & x \in D \setminus \{x_i\} \end{cases}$
 "Lagrange polynomial for x_i over D "

Hmm when $n=K$, the encoding function for RS code
is JUST a change of basis..!

In fact, RS is a linear code, so for specified params

s.t. $\text{Enc}: \mathbb{F}_q^K \rightarrow \mathbb{F}_q^N$

$$m \mapsto m \cdot G$$

$$\exists G: \mathbb{F}_q^{K \times n}$$



G is the change of basis matrix.

Typically $\square \cdot \square$ takes $\Theta(Kn) = \Theta(n^2)$... but we have a nice \boxed{G} with great structure embedded.

We can equivalently do $\text{Enc}: \mathbb{F}_q^k \rightarrow \mathbb{F}_q^n$

$m \mapsto$ change of basis to m

and when our bases are chosen well, this will be $O(n \log n)$ 

For us, $\mathbb{F}_q = \mathbb{F}_{2^{2^L}} = \mathbb{T}_L$ // binary field

So our msg will be K coeffs w.r.t. Novel Poly Basis
 and our codeword will be K coeffs w.r.t. Lagrange Basis Coeffs
 over $D \subseteq \mathbb{F}_q$, $|D| = K$

What about when $K \neq n$?

We will support K, n as powers of 2 only, $n < 2^{2^L}$

let's say that $n = 2^R \cdot K$ $\nearrow R = \log\left(\frac{n}{K}\right)$
 (here $R = \log\text{-inv-rate}$)

We will let $D = \{w_b \mid b \in \{0, \dots, n-1\}\} \subseteq \mathbb{T}_L$

Interestingly, NOT only is $D \subseteq T_L$, but it's also a subspace. $D = \langle \beta_0, \dots, \beta_{\lg(n)-1} \rangle$

D can be partitioned into 2^R equally sized subsets

$$D_0 = \langle \beta_0, \dots, \beta_{\lg(K)-1} \rangle = w_0 + D_0 \quad \text{called "cosets"}$$

$$D_1 = \{ w_K + d \mid d \in D_0 \} = w_K + D_0 \quad \text{ask Ben/Raju for more!}$$

$$D_2 = \{ w_{2K} + d \mid d \in D_0 \} = w_{2K} + D_0$$

$$\dots$$

$$D_i = \{ w_{ik} + d \mid d \in D_0 \} = w_{ik} + D_0$$

$$\dots$$

$$D_{2^R-1} = \{ w_{(2^R-1)K} + d \mid d \in D_0 \} = w_{(2^R-1)K} + D_0$$

$$w_K = \beta_{\lg(K)}$$

$$w_{2K} = \beta_{\lg(K)+1}$$

$$w_{3K} = \beta_{\lg(K)} + \beta_{\lg(K)+1}$$

- Note that • $D = D_0 \cup D_1 \cup \dots \cup D_{2^R-1}$
- $\forall 0 \leq i < j \leq 2^R-1, D_i \cap D_j = \emptyset$
- $\forall i, |D_i| = |D_0| = K$

So, what does our RS over binary field encoding algo look like now?

Input: msg $\vec{m} \in T_K^K$, interpreted as K coeffs wrt Novel Poly Basis

Algo:

$$\text{codeword} = []$$

for coset-shift in $\{0, \dots, 2^R - 1\}$:

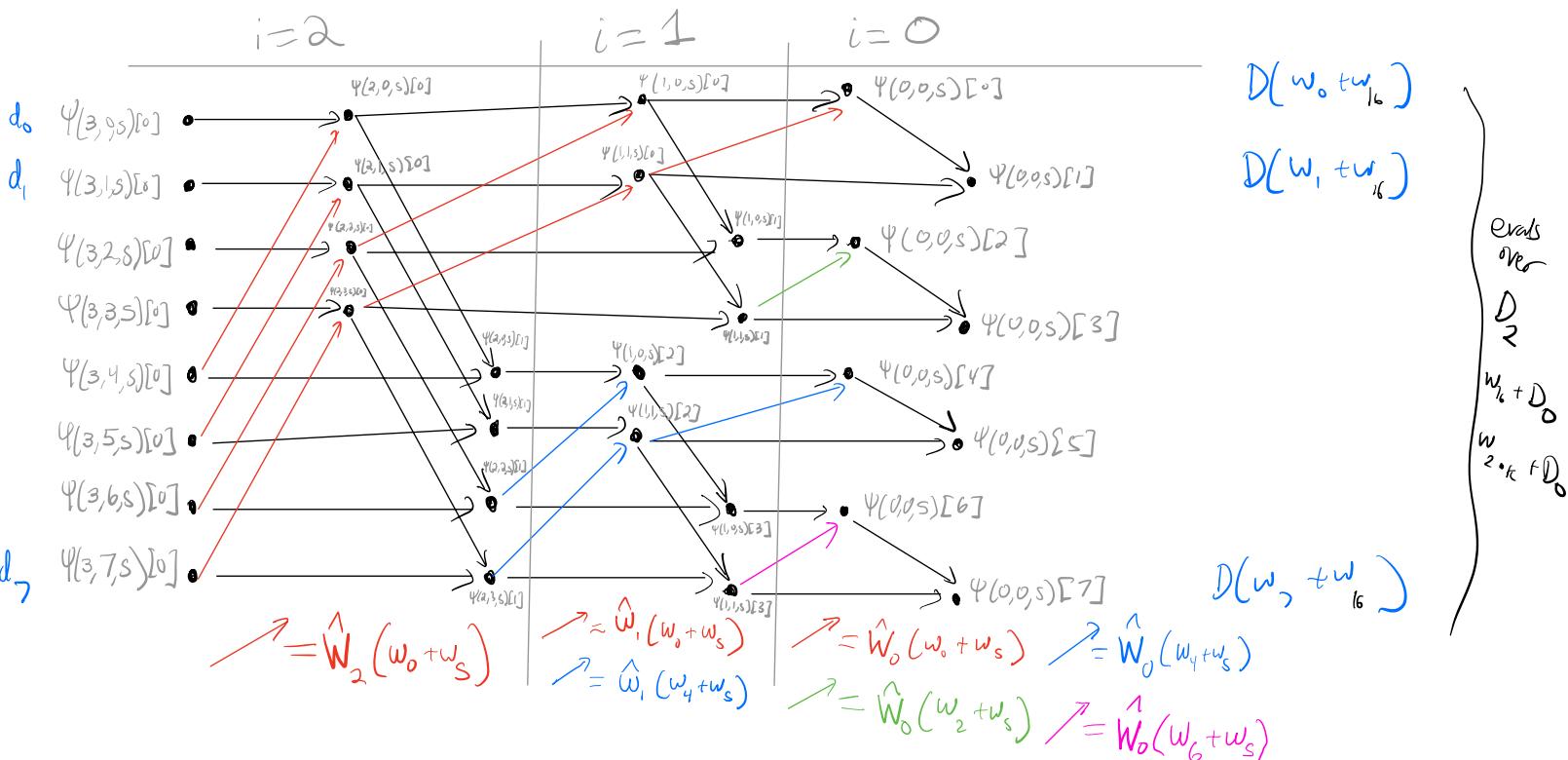
- chunk = AdditiveNTT(\vec{m} , $s = \text{coset_shift} * K$)
- extend codeword with chunk

return codeword

Runtime: $\Theta(K \log K \cdot 2^R) = \Theta(n \log K)$

\rightarrow Zooming in to AdditiveNTT(m, s)

- //
- Forward transform when $\ell = \lg(k)$, $k=8$, arbitrary s
 - $\vec{m} = [\Psi(3, 0, s)[0], \dots, \Psi(3, 7, s)[0]]$



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Quick note on inverse transform and decoding

- You can decode from any chunk using iNTT
Recall the i th chunk of size K is Lagrange Basis coeffs wrt D_i
- The inverse transform follows the exact same recursive structure just starting from

$$c = \left[\underbrace{\Psi(0, 0, s)[0], \dots, \Psi(0, 0, s)[K-1]}_{\text{. . .}} \right]$$

$\// s = K \cdot 2^i$

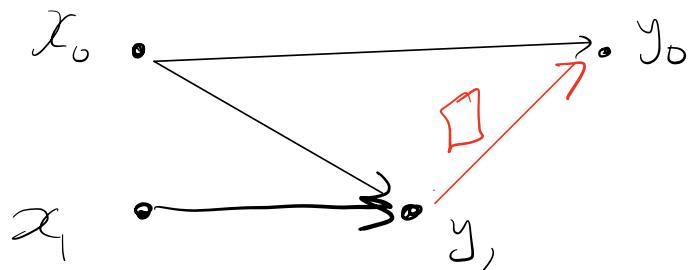
to

$$m = \left[\Psi(\lg(k), 0, s)[0], \dots, \Psi(\lg(k), K-1, s)[0] \right]$$

Appendix:

Additive NTT

Structure



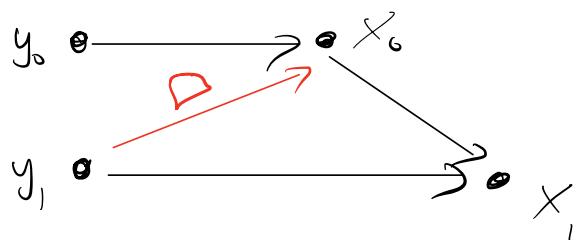
$$y_1 = x_0 + x_1$$

$$y_0 = x_0 + \square y_1$$

$$(x_0, x_1) \mapsto (\square x_0 + \square x_1 + x_0, x_0 + x_1)$$

Additive NTT

Structure



$$x_0 = y_0 + \square y_1$$

$$x_1 = x_0 + y_1$$

$$(y_0, y_1) \rightarrow (y_0 + \square y_1, y_0 + \square y_1 + y_1)$$

Composition, in characteristic 2, will be identity map