

# Decoding Downset Codes over Finite Grids

S. Venkitesh

Department of Mathematics, IIT Bombay

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# Introduction

Field  $\mathbb{F}$  (arbitrary)

Finite grid (nonempty, finite set)

$$S = S_1 \times \cdots \times S_n = \{p_1, \dots, p_k\} \subseteq \mathbb{F}^n$$

$$f : S \rightarrow \mathbb{F}$$



$$f = (f(p_1), \dots, f(p_k))$$

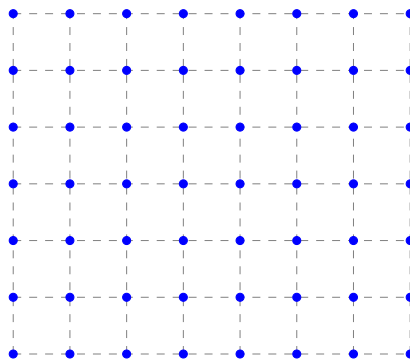


$$f = P : S \rightarrow \mathbb{F}$$

(polynomial function)

$$\deg_{X_i} P(\mathbf{X}) \leq |S_i| - 1, \forall i \in [n]$$

$$|S_2| - 1$$



$$0$$

$$|S_1| - 1$$

$$\mathcal{M}_S = \{\mathbf{X}^\alpha = X_1^{\alpha_1} \cdots X_n^{\alpha_n} : \alpha_i \leq |S_i| - 1, \forall i \in [n]\}$$

$$\{S \rightarrow \mathbb{F}\} \simeq \text{span}_{\mathbb{F}} \mathcal{M}_S$$

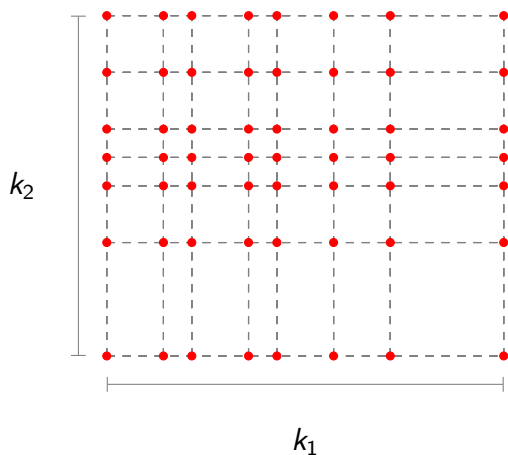
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# Introduction

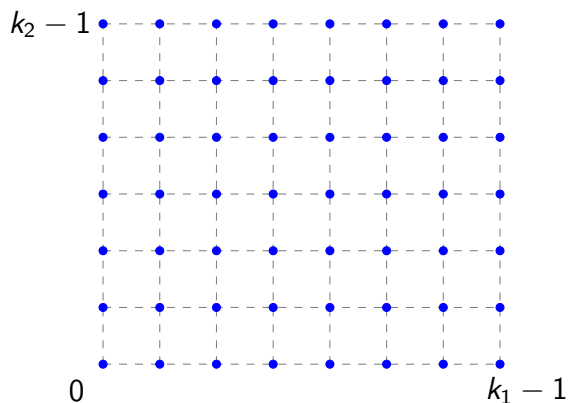
$$S = S_1 \times \cdots \times S_n \subseteq \mathbb{F}^n, \quad k_i = |S_i|, \quad \forall i \in [n], \quad k = |S| = k_1 \cdots k_n$$

$$\mathcal{M}_S = \{\mathbf{X}^\alpha = X_1^{\alpha_1} \cdots X_n^{\alpha_n} : \alpha_i \leq k_i - 1, \forall i \in [n]\} = \mathbb{F}\text{-basis of } \{S \rightarrow \mathbb{F}\}$$

$$S \subseteq \mathbb{F}^2$$



$$\mathcal{M}_S \subseteq \mathbb{N}^2$$



# Introduction

## Definition (Reed-Muller Code on $S$ )

$$RM(S, d) = \{P(\mathbf{X}) \in \text{span}_{\mathbb{F}}\mathcal{M}_S \mid \deg P \leq d\}, \quad d \leq \sum_{i=1}^n (k_i - 1)$$

$RM(S, d)$  is a subspace of  $\text{span}_{\mathbb{F}}\mathcal{M}_S \simeq \mathbb{F}^k$ , i.e. it is a **linear code**.

$$RM(S, d) = \text{span}_{\mathbb{F}}\mathcal{M}_{S,d}, \quad \text{where } \mathcal{M}_{S,d} = \left\{ \mathbf{X}^\alpha \in \mathcal{M}_S : |\alpha| = \sum_{i=1}^n \alpha_i \leq d \right\}.$$

Note that

$$\mathbf{X}^\alpha \in \mathcal{M}_{S,d}, \mathbf{X}^\beta \mid \mathbf{X}^\alpha \text{ (i.e. } \beta \leq \alpha \text{ in natural partial order)} \implies \mathbf{X}^\beta \in \mathcal{M}_{S,d}.$$

So  $\mathcal{M}_{S,d}$  is a **down-closed set (downset)**.

# Downset and Downset Code

## Definition (Downset)

A nonempty set of monomials  $\mathcal{D}$  is called a **downset** if

$$\mathbf{x}^\alpha \in \mathcal{D}, \mathbf{x}^\beta \mid \mathbf{x}^\alpha \ (\beta \leq \alpha) \implies \mathbf{x}^\beta \in \mathcal{D}.$$

## Definition (Downset code on $S$ )

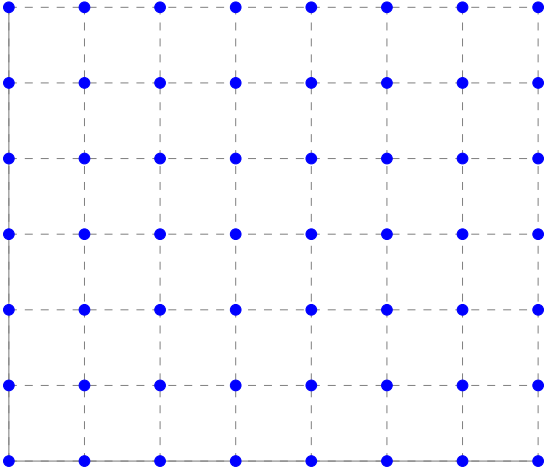
$$\mathcal{C}(S, \mathcal{D}) = \text{span}_{\mathbb{F}} \mathcal{D}, \quad \text{where } \mathcal{D} \subseteq \mathcal{M}_S \text{ is a downset}$$

$\mathcal{C}(S, \mathcal{D})$  is a subspace of  $\text{span}_{\mathbb{F}} \mathcal{M}_S \simeq \mathbb{F}^k$ , i.e., it is a linear code.

The downset  $\mathcal{D}$  is an  $\mathbb{F}$ -basis of  $\mathcal{C}(S, \mathcal{D})$ .

# Examples of Downsets and Downset codes

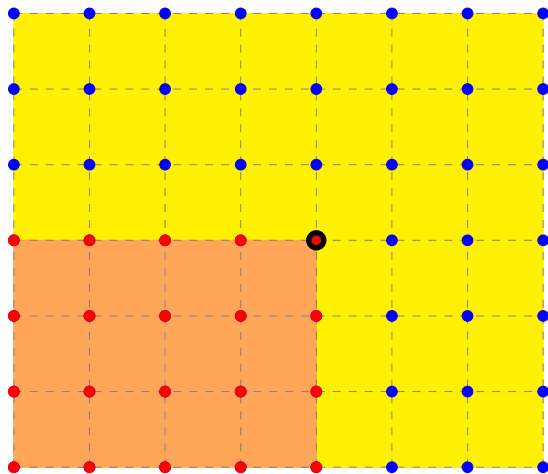
**Eg.**  $n = 2$ ,  $S = S_1 \times S_2$ ,  $|S_1| = 8$ ,  $|S_2| = 7$ ,  $\mathcal{M}_S = \{0, \dots, 7\} \times \{0, \dots, 6\}$



$\mathcal{M}_S$

## Examples of Downsets and Downset codes

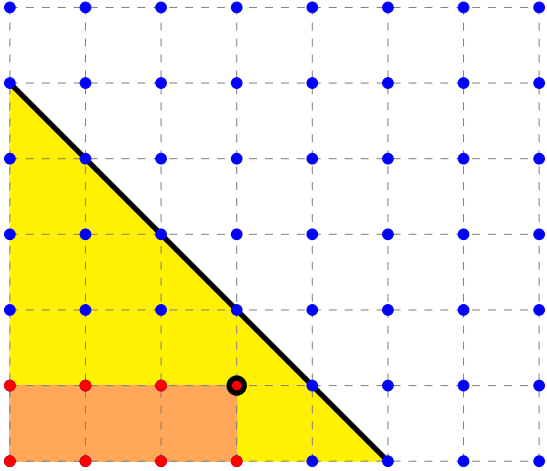
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$$\mathcal{D} = \mathcal{M}_S, \quad \mathcal{C}(S, \mathcal{D}) = \{S \rightarrow \mathbb{F}\}$$

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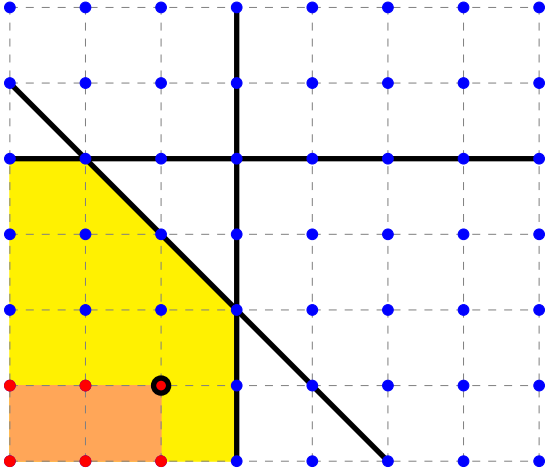


$$\mathcal{D} = \{\mathbf{x}^\alpha \in \mathcal{M}_S : |\alpha| \leq 5\}, \quad \mathcal{C}(S, \mathcal{D}) = RM(S, 5)$$



# Examples of Downsets and Downset codes

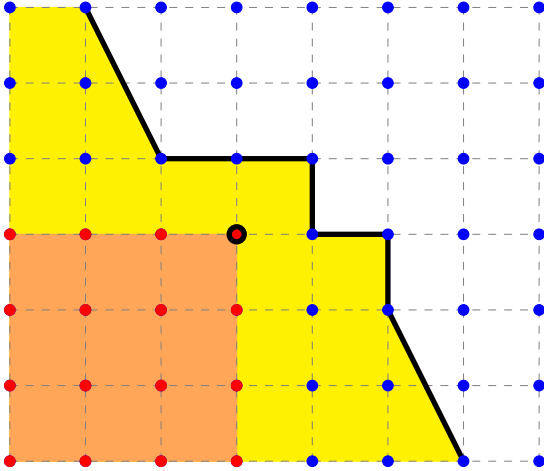
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$$\mathcal{D} = \{\mathbf{x}^\alpha \in \mathcal{M}_S : |\alpha| \leq 5, \alpha_1 \leq 3, \alpha_2 \leq 4\}, \quad \mathcal{C}(S, \mathcal{D})$$

# Examples of Downsets and Downset codes

**Eg.**  $n = 2$ ,  $S = S_1 \times S_2$ ,  $|S_1| = 8$ ,  $|S_2| = 7$ ,  $\mathcal{M}_S = \{0, \dots, 7\} \times \{0, \dots, 6\}$



$\mathcal{D}, \mathcal{C}(S, \mathcal{D})$

## What is (unique) decoding?

Linear code  $\mathcal{C}$

Definition (Hamming distance (metric))

$$\Delta(f, g) = |\{x \in S : f(x) \neq g(x)\}|, \quad f, g \in \mathcal{C}$$

The **Hamming weight** of  $f \in \mathcal{C}$  is  $\|f\| = |\text{supp}(f)| = \Delta(f, 0)$ .

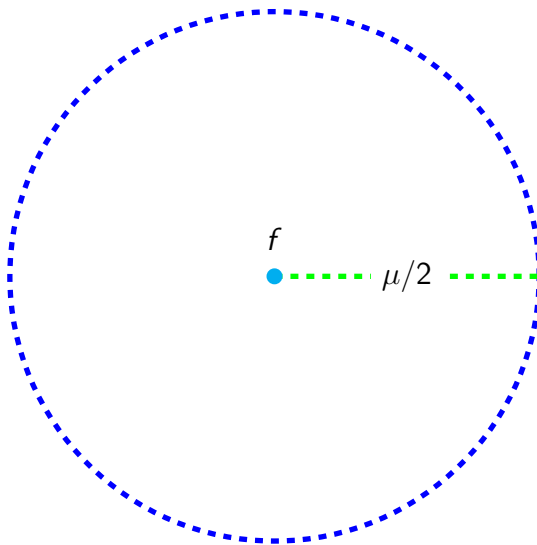
Definition (Minimum distance of a linear code)

$$\mu(\mathcal{C}) = \min\{\Delta(f, g) : f, g \in \mathcal{C}, f \neq g\} = \min\{\|f\| : f \in \mathcal{C}, f \neq 0\}$$

## What is (unique) decoding?

Linear code  $C \subseteq \mathbb{F}^k$ ,  $\mu = \mu(C)$ ,  $f \in \mathbb{F}^k$

$$B(f, \mu/2) = \{g \in \mathbb{F}^k : \Delta(f, g) < \mu/2\}$$

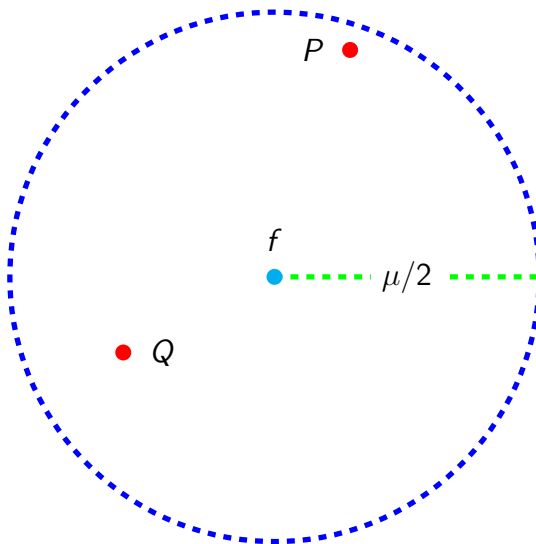


Case (1)  $B(f, \mu/2)$  with  $B(f, \mu/2) \cap C = \emptyset$ , **NO DECODING**

# What is (unique) decoding?

Linear code  $\mathcal{C} \subseteq \mathbb{F}^k$ ,  $\mu = \mu(\mathcal{C})$ ,  $f \in \mathbb{F}^k$ ,  $P, Q \in \mathcal{C}$ ,  $P \neq Q$

$$B(f, \mu/2) = \{g \in \mathbb{F}^k : \Delta(f, g) < \mu/2\}$$



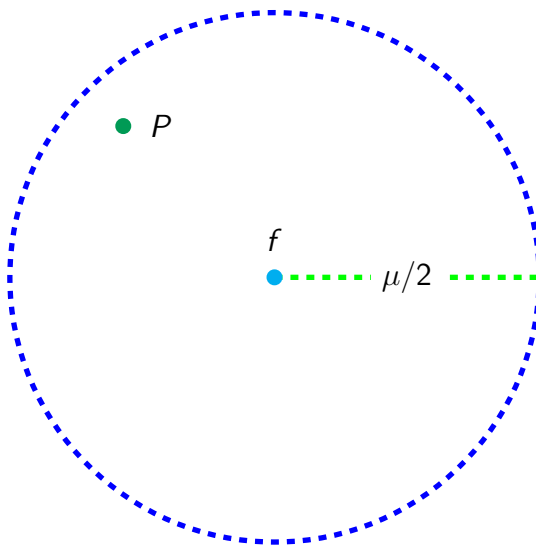
Case (II)  $B(f, \mu/2)$  with  $B(f, \mu/2) \cap \mathcal{C} \supseteq \{P, Q\}$ , **NOT POSSIBLE**

$$\Delta(P, Q) \leq \Delta(f, P) + \Delta(f, Q) < \mu/2 + \mu/2 = \mu \quad \text{NOT TRUE}$$

## What is (unique) decoding?

Linear code  $\mathcal{C} \subseteq \mathbb{F}^k$ ,  $\mu = \mu(\mathcal{C})$ ,  $f \in \mathbb{F}^k$ ,  $P, Q \in \mathcal{C}$ ,  $P \neq Q$

$$B(f, \mu/2) = \{g \in \mathbb{F}^k : \Delta(f, g) < \mu/2\}$$



Case (III)  $B(f, \mu/2)$  with  $B(f, \mu/2) \cap \mathcal{C} = \{P\}$ , **UNIQUE DECODING**

## Theorem (A template)

For 'appropriate' dimension  $n$ , field  $\mathbb{F}$ , finite grid  $S \subseteq \mathbb{F}^n$  and the corresponding linear code  $\mathcal{C} \subseteq \{S \rightarrow \mathbb{F}\}$ , there is an algorithm which, given  $f : S \rightarrow \mathbb{F}$  with the 'promise' that there exists a (unique)  $P \in \mathcal{C}$  such that  $\Delta(f, P) < \mu(\mathcal{C})/2$ , returns  $P$  'efficiently'.

	$n$	$\mathbb{F}$	$S$	$\mathcal{D}$	$\mathcal{C}(S, \mathcal{D})$
Reed (1954)	arbitrary	$\mathbb{F}_2$	$\mathbb{F}_2^n$	$\mathcal{M}_{S,d}$	$RM(\mathbb{F}_2^n, d)$
	arbitrary	arbitrary	$\{0, 1\}^n$	$\mathcal{M}_{S,d}$	$RM(\{0, 1\}^n, d)$
Forney (1966) <sup>1</sup>	1	$\mathbb{F}_q$	$\mathbb{F}_q$	$\mathcal{M}_{S,d}$	$RS(\mathbb{F}_q, d)$
	1	arbitrary	arbitrary	$\mathcal{M}_{S,d}$	$RS(S, d)$
Berlekamp, Welch (1983)	1	$\mathbb{F}_q$	$\mathbb{F}_q$	$\mathcal{M}_{S,d}$	$RS(\mathbb{F}_q, d)$
	1	$\mathbb{F}_q$	arbitrary	$\mathcal{M}_{S,d}$	$RS(\mathbb{F}_q, d)$
Kim, Kopparty (2017)	arbitrary	arbitrary	arbitrary	$\mathcal{M}_{S,d}$	$RM(S, d)$
Our result <sup>2</sup>	arbitrary	arbitrary	arbitrary	arbitrary	$\mathcal{C}(S, \mathcal{D})$

<sup>1</sup>with weights/uncertainties on words.

<sup>2</sup>joint work with Srikanth Srinivasan and Utkarsh Tripathi, both from Dept. of Mathematics, IIT Bombay.

# Main Theorem

## Theorem (Our Result)

*There is a deterministic polynomial time algorithm such that, given a finite grid  $S = S_1 \times \cdots \times S_n \subseteq \mathbb{F}^n$ , a downset  $\mathcal{D} \subseteq \mathcal{M}_S$ , and  $f : S \rightarrow \mathbb{F}$ , the algorithm outputs  $C \in \mathcal{C}(S, \mathcal{D})$  such that  $\Delta(f, C) < \frac{\mu(S, \mathcal{D})}{2}$ , if such a  $C$  exists. If such a  $C$  does not exist, then the algorithm outputs an arbitrary polynomial.*

We will prove a slightly stronger version of the above involving a **weighted word** as input.



# Weighted word and weighted distance

## Definition

Word  $b : S \rightarrow \mathbb{F}$

Weighted word  $(a, w) : S \rightarrow \mathbb{F} \times [0, 1]$

Weighted distance  $\Delta((a, w), b) = \sum_{a(x)=b(x)} \left( \frac{w(x)}{2} \right) + \sum_{a(x) \neq b(x)} \left( 1 - \frac{w(x)}{2} \right)$

## Fact (Triangle Inequality)

Let  $(a, w) : S \rightarrow \mathbb{F} \times [0, 1]$  be a weighted word and  $b, c : S \rightarrow \mathbb{F}$  be words. Then

$$\Delta((a, w), b) + \Delta((a, w), c) \geq \Delta(b, c).$$

Further if  $b, c \in \mathcal{C}(S, \mathcal{D})$  and  $b \neq c$ , then

$$\Delta((a, w), b) + \Delta((a, w), c) \geq \Delta(b, c) \geq \mu(S, \mathcal{D}).$$

In particular,  $\Delta((a, w), b) < \mu(S, \mathcal{D})/2$  and  $\Delta((a, w), c) < \mu(S, \mathcal{D})/2$  is not possible.

## Main Theorem (Weighted version)

### Theorem (Our Result)

*There is a deterministic polynomial time algorithm such that, given a finite grid  $S = S_1 \times \cdots \times S_n \subseteq \mathbb{F}^n$ , a downset  $\mathcal{D} \subseteq \mathcal{M}_S$ , and  $f : S \rightarrow \mathbb{F}$ , the algorithm outputs  $C \in \mathcal{C}(S, \mathcal{D})$  such that  $\Delta(f, C) < \frac{\mu(S, \mathcal{D})}{2}$ , if such a  $C$  exists. If such a  $C$  does not exist, then the algorithm outputs an arbitrary polynomial.*

### Theorem (Our Result, weighted version)

*There is a deterministic polynomial time algorithm such that, given a finite grid  $S = S_1 \times \cdots \times S_n \subseteq \mathbb{F}^n$ , a downset  $\mathcal{D} \subseteq \mathcal{M}_S$ , and a **weighted word**  $(a, w) : S \rightarrow \mathbb{F}$ , the algorithm outputs  $C \in \mathcal{C}(S, \mathcal{D})$  such that  $\Delta((a, w), C) < \frac{\mu(S, \mathcal{D})}{2}$ , if such a  $C$  exists. If such a  $C$  does not exist, then the algorithm outputs an arbitrary polynomial.*

We will proceed by induction on  $n$ . The base case uses Forney's weighted Reed-Solomon decoder, mentioned earlier. For this talk, this decoder is a BLACK BOX.

## Some more facts

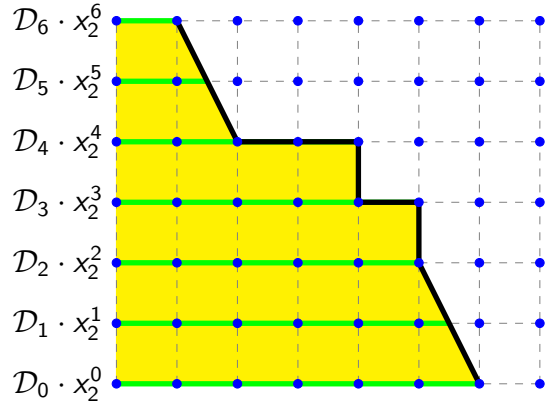
- Let  $\mathcal{D} \subseteq \mathcal{M}_S$  be a downset and  $\deg_n \mathcal{D} = \max\{\alpha_n : \alpha \in \mathcal{D}\}$ . Define

$$\mathcal{D}_i = \{\beta \in \mathcal{M}_S : (\beta, i) \in \mathcal{D}\},$$

for  $i \in \{0, \dots, d = \deg_n \mathcal{D}\}$ .

Then

- (i)  $\mathcal{D}_i$  is a downset, for all  $i$ , since  $\mathcal{D}$  is a downset.
- (ii)  $\mathcal{D}_0 \supseteq \dots \supseteq \mathcal{D}_d$ .



- Let  $\tilde{S} = S_1 \times \dots \times S_{n-1}$ . For every  $i \in \{0, \dots, d = \deg_n \mathcal{D}\}$ , we have

$$\mu(S, \mathcal{D}) \leq \mu(\tilde{S}, \mathcal{D}_i) \cdot \mu(S_n, \{0, \dots, i\}).$$

## The Decoding Algorithm (An outline)

**Base Case.**  $n = 1$ . In this case, the algorithm is Forney's weighted decoding algorithm.

**Induction Hypothesis.** The weighted decoding algorithm works in the case  $n = 2$ .

Suppose  $n = 3$ .

**Input:**  $(S_1 \times S_2 \times S_3, \mathcal{D}, (a, w))$ , the grid, the downset and the weighted word.

Let  $d = \deg_3 \mathcal{D}$ .

Let the correct codeword be

$$C(x, y, z) = \sum_{j=0}^d Q_j(x, y)z^j, \quad \forall (x, y, z) \in S_1 \times S_2 \times S_3.$$

It is then enough to find  $Q_j : S_1 \times S_2 \rightarrow \mathbb{F}$ ,  $j = 0, \dots, d$ . We will run  $i \searrow d, \dots, 0$ .

It is here, in finding  $Q_j$ -s that we use the inductive hypothesis, since the  $Q_j$ -s are 2-variate. In order to do this, we need to find suitable weighted words.

## The Decoding Algorithm (An outline)

Now consider a fixed  $i$  and suppose that at the  $i$ -th stage, the functions  $Q_j(x, y)$ ,  $j \searrow d, \dots, i + 1$  are known. When  $i = d$ , nothing is known (that's fine!). Let

$$a_i(x, y, z) = a(x, y, z) - \sum_{j=i+1}^d Q_j(x, y)z^j.$$

For every  $(x, y) \in S_1 \times S_2$ , define

$$a_{i,(x,y)}(z) = a_i(x, y, z), \quad w_{(x,y)}(z) = w(x, y, z).$$

Then we use Forney's algorithm for the one variable case. Apply the algorithm on the input  $(S_3, \{0, \dots, i\}, a_{i,(x,y)})$  to get the 'possibly correct 1-variate' word  $G_{i,(x,y)} : S_3 \rightarrow \mathbb{F}$ .

We have thus computed a 1-variate word  $G_{i,(x,y)} : S_3 \rightarrow \mathbb{F}$  for each  $(x, y) \in S_1 \times S_2$ .

## The Decoding Algorithm (An outline)

We now 'compare' weighted distances and determine the 'input' weighted words to be passed on to the (correct) 2-variable weighted decoder. Let

$$\Delta_{i,(x,y)} = \Delta(a_{i,(x,y)}, G_{i,(x,y)}), \quad \mu_i = \frac{\mu(S_n, \{0, \dots, i\})}{2}.$$

If  $\Delta_{i,(x,y)} < \mu_i$ , let  $\sigma_i(x, y) = [z^i](G_{i,(x,y)})$ ,  $\delta_i(x, y) = \frac{\Delta_{i,(x,y)}}{\mu_i}$ .

If  $\Delta_{i,(x,y)} \geq \mu_i$ , let  $\sigma_i(x, y) = 0$ ,  $\delta_i(x, y) = 1$ .

We have thus computed a 2-variate weighted word  $(\sigma_i, \delta_i) : S_1 \times S_2 \rightarrow \mathbb{F}$ . We then give the input  $(S_1 \times S_2, \mathcal{D}_i, (\sigma_i, \delta_i))$  to the 2-variate weighted decoder to get the output function  $P_i : S_1 \times S_2 \rightarrow \mathbb{F}$ .

Since by inductive assumption, the algorithm is correct for the 2-variate case, we have  $P_i = Q_i$ . (It is a routine case analysis to check that  $\Delta(P_i, Q_i) < \mu(S_1 \times S_2, \mathcal{D}_i)/2$ . This is why the correct decoder gives the correct output!)

Running through  $i \searrow d, \dots, 0$  gives the correct codeword  $C(x, y, z) = \sum_{j=0}^d Q_j(x, y)z^j$ .

# Thank You!