

# On Average Baby PIH and Its Applications

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# Constraint Satisfaction Problem ( $q$ CSP)

- Variables  $X = \{x_1, \dots, x_n\}$
- Alphabet  $\Sigma$
- Constraints  $\Phi = \{\varphi_1, \dots, \varphi_m\}$ , each depends on  $q$  variables
- Decide: whether it's satisfiable or not.

**NP-Complete.**

# The PCP Theorem [AS-ALMSS'98] [Dinur'07]

- **NP-hard** to decide whether a  $q$ CSP instance is
  - Satisfiable, or
  - Cannot satisfy  $s$ -fraction of constraints simultaneously.  
 $(0 < s < 1)$

# Relaxation: Multi-Assignment

- Assign each variable a **set** of values.

$$\begin{array}{ll} x_1: \{ 1, 5, 7, 9 \} & \varphi_1 = (x_1 x_2, C_1) \\ x_2: \{ 2, 3, 4 \} & \varphi_2 = (x_2 x_3, C_2) \\ x_3: \{ 2, 6 \} & \varphi_3 = (x_2 x_4, C_3) \\ x_4: \{ 4, 5, 6, 8 \} & \end{array}$$

The diagram illustrates the multi-assignment of variables. It shows four sets of values for variables  $x_1$ ,  $x_2$ ,  $x_3$ , and  $x_4$ . The first set  $x_1$  contains {1, 5, 7, 9}, with 1 highlighted. The second set  $x_2$  contains {2, 3, 4}, with 2 highlighted. The third set  $x_3$  contains {2, 6}, with 2 highlighted. The fourth set  $x_4$  contains {4, 5, 6, 8}, with no values highlighted. Three yellow arrows point from the highlighted values to their respective constraints:  $\varphi_1 = (x_1 x_2, C_1)$ ,  $\varphi_2 = (x_2 x_3, C_2)$ , and  $\varphi_3 = (x_2 x_4, C_3)$ .

# Relaxation: Multi-Assignment

- Assign each variable a **set** of values.

$x_1: \{ 1, 5, 7, 9 \}$

$\varphi_1 = (x_1 x_2, C_1)$

$x_2: \{ 2, 3, 4 \}$

$\varphi_2 = (x_2 x_3, C_2)$

$x_3: \{ 2, 6 \}$

$\varphi_3 = (x_2 x_4, C_3)$

$x_4: \{ 4, 5, 6, 8 \}$



# Relaxation: Multi-Assignment

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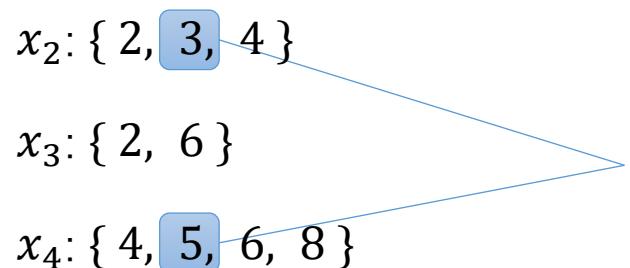
$x_2: \{ 2, 3, 4 \}$

$\varphi_2 = (x_2 x_3, C_2)$

$x_3: \{ 2, 6 \}$

$\varphi_3 = (x_2 x_4, C_3)$

$x_4: \{ 4, 5, 6, 8 \}$



# Multi-Assignment PCP [Arora,Moshkovitz,Safra'06]

- **NP-hard** to decide whether a  $q$ CSP instance is
  - Satisfiable, or
  - Cannot satisfy  $s$ -fraction of constraints simultaneously **even** when **each** variable assigned  $\leq t$  values.  
 $(0 < s < 1, \ t > 1)$
- Used to prove NP-hardness of approximating SetCover.

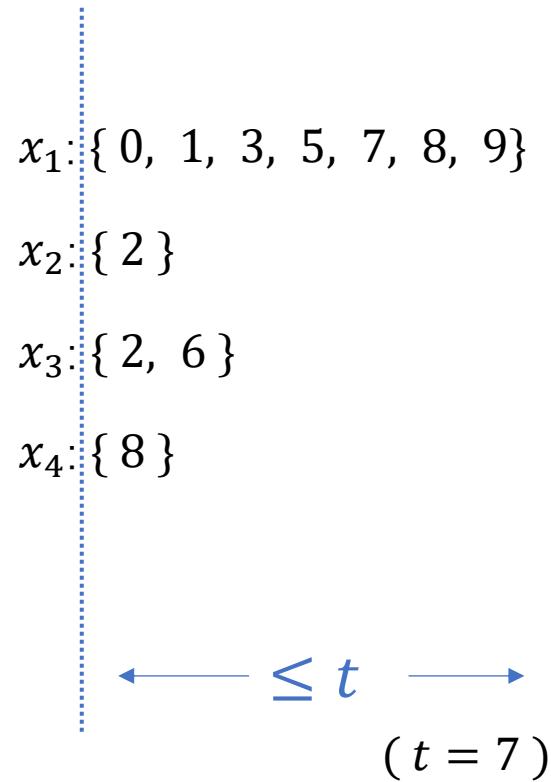
# Parameterized Inapprox. Hypo. (PIH)

- Hypothesis [Lokshtanov, Ramanujan, Saurabh, Zehavi'20]:  
No FPT algorithm decide a 2CSP parameterized by  $k = |X|$  is:
  - Satisfiable, or
  - Cannot satisfy  **$s$ -fraction** of constraints simultaneously.  $(0 < s < 1)$
- SOTA: Exponential Time Hypothesis  $\rightarrow$  PIH. [Guruswami, Lin, Ren, Sun, Wu'24]
- Major open problem:  $\text{W}[1] \neq \text{FPT} \rightarrow \text{PIH} ?$

# Weaken: Baby PIH [Guruswami,Ren,Sandeep'24]

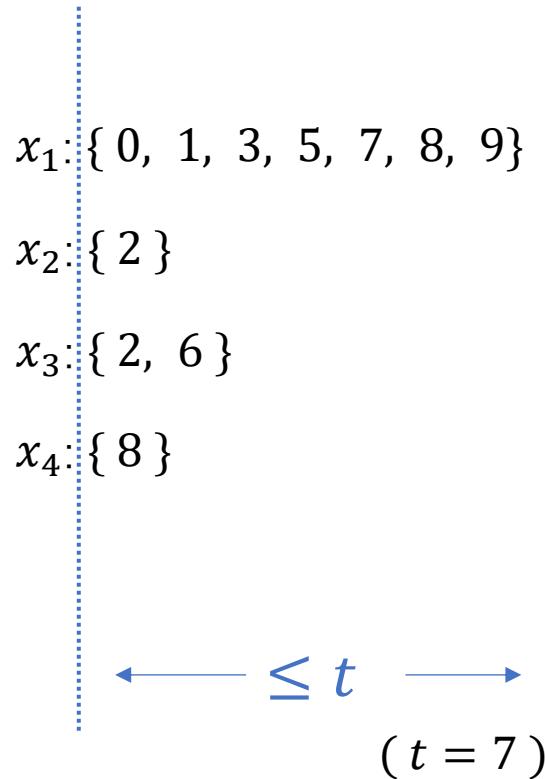
- No FPT algorithm for deciding a 2CSP parameterized by  $k = |X|$ :
  - Being satisfiable, or
  - Cannot satisfy all constraints simultaneously even when **each** variable assigned  $\leq t$  values. ( $t > 1$ )
- $\text{W[1]} \neq \text{FPT} \rightarrow$  Baby PIH. [Guruswami,Ren,Sandeep'24]
  - Following the method in [Barto,Kozik'22] showing Baby PCP without using PCP Theorem.

# Weaken: Baby PIH [Guruswami,Ren,Sandeep'24]



# Question: Average Baby PIH

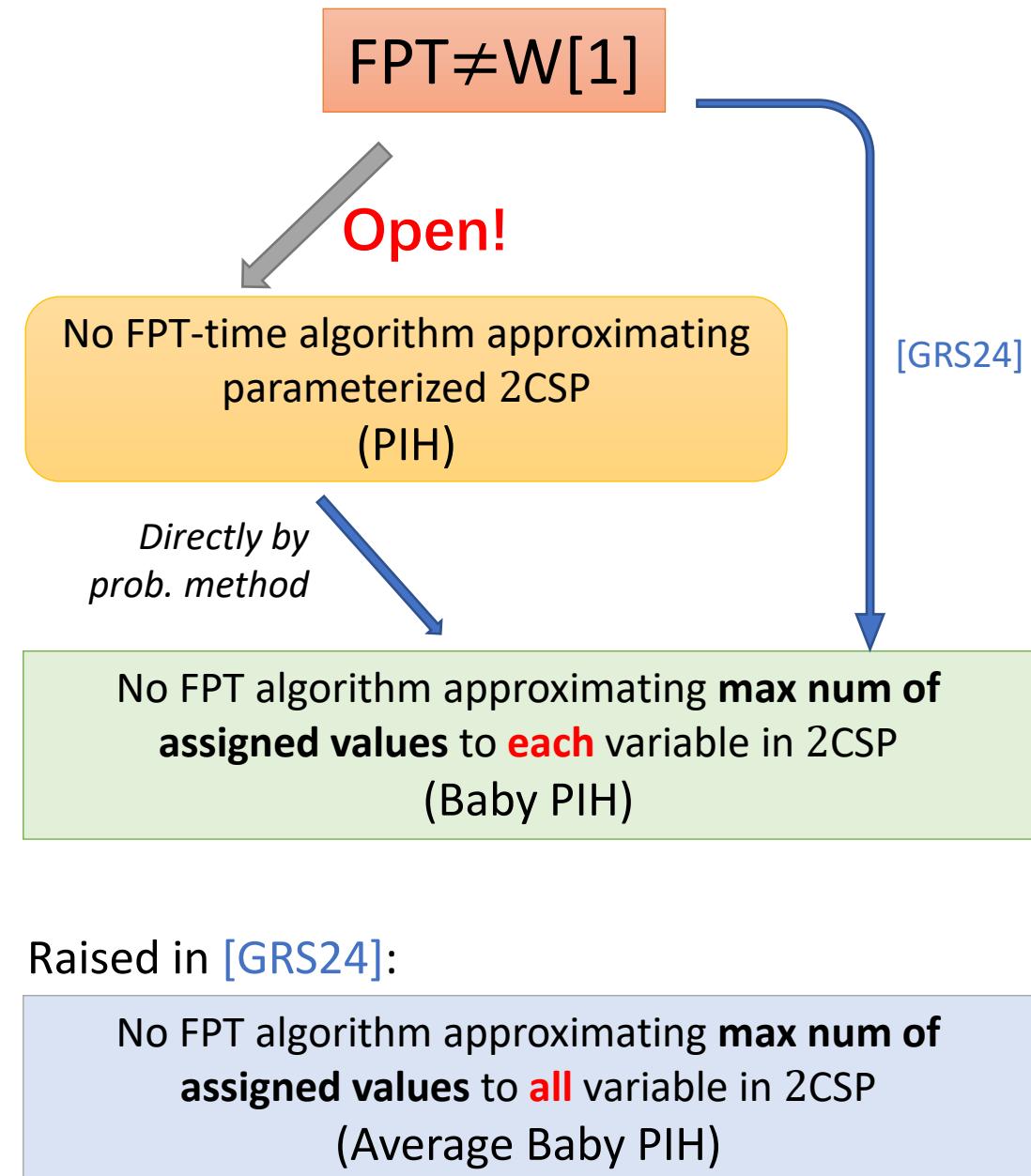
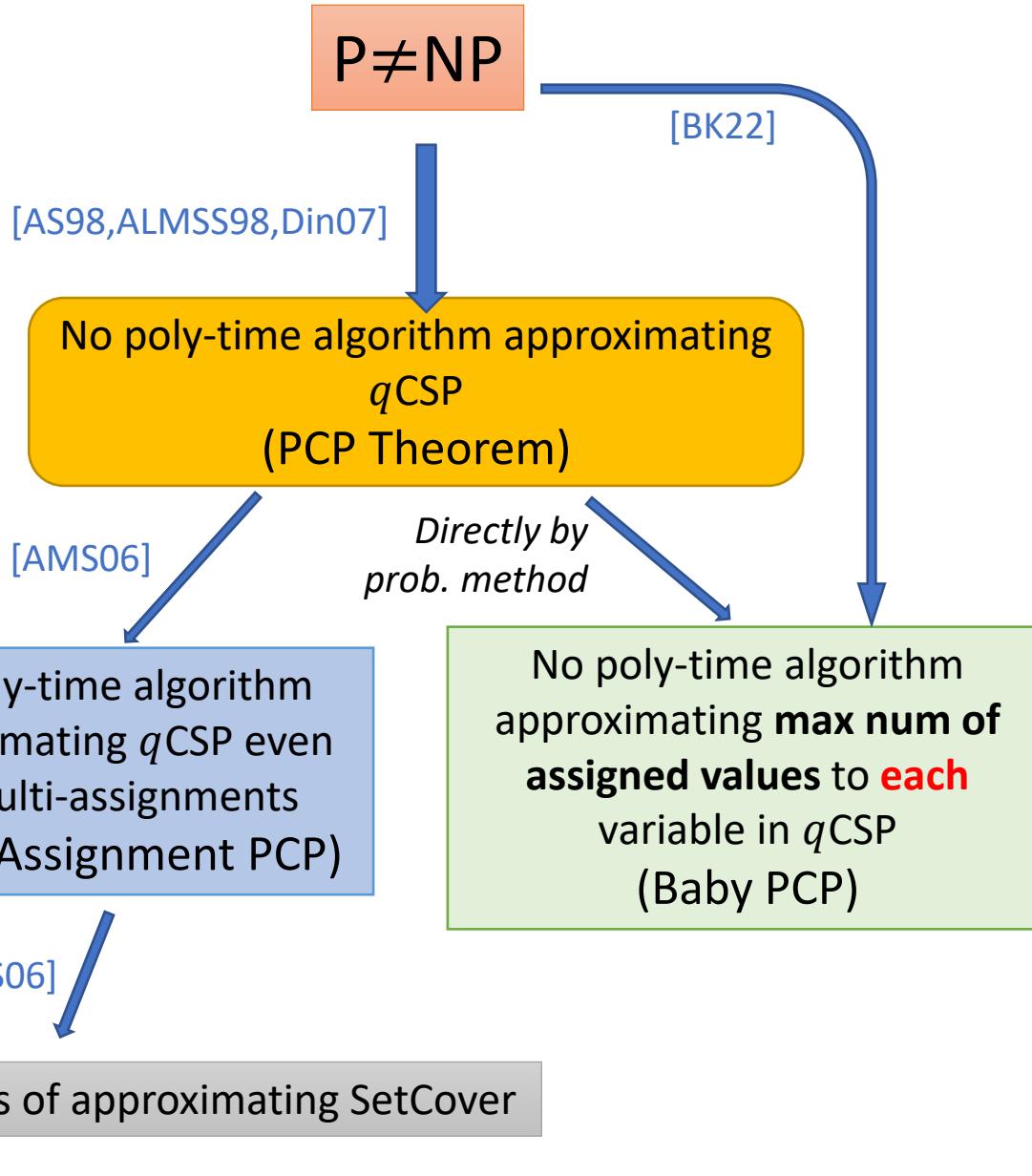
$$|X| = 4,$$



$$\text{Total \# of values: } 7 + 1 + 2 + 1 = 11 = 2.75|X|.$$

# Question: Average Baby PIH

- No FPT algorithm for deciding a 2CSP parameterized by  $k = |X|$ :
  - Being satisfiable, or
  - Cannot satisfy all constraints simultaneously even when assigning to  $X$  less than  $t|X|$  values **in total**.  $(t > 1)$ 
    - $\ell_1$  instead of  $\ell_\infty$
- Raised in [Guruswami,Ren,Sandeep'24].



# Our result

$W[1] \neq FPT$   Average Baby PIH

$P \neq NP$

[AS98, ALMSS98, Din07]

No poly-time algorithm approximating  
 $q$ CSP  
(PCP Theorem)

[AMS06]

No poly-time algorithm  
approximating  $q$ CSP even  
for multi-assignments  
(Multi-Assignment PCP)

[AMS06]

Hardness of approximating SetCover

[BK22]

*Directly by  
prob. method*

No poly-time algorithm  
approximating **max num of  
assigned values to each**  
variable in  $q$ CSP  
(Baby PCP)

$FPT \neq W[1]$

**Open!**

No FPT-time algorithm approximating  
parameterized 2CSP  
(PIH)

*Directly by  
prob. method*

No FPT algorithm approximating **max num of  
assigned values to each** variable in 2CSP  
(Baby PIH)

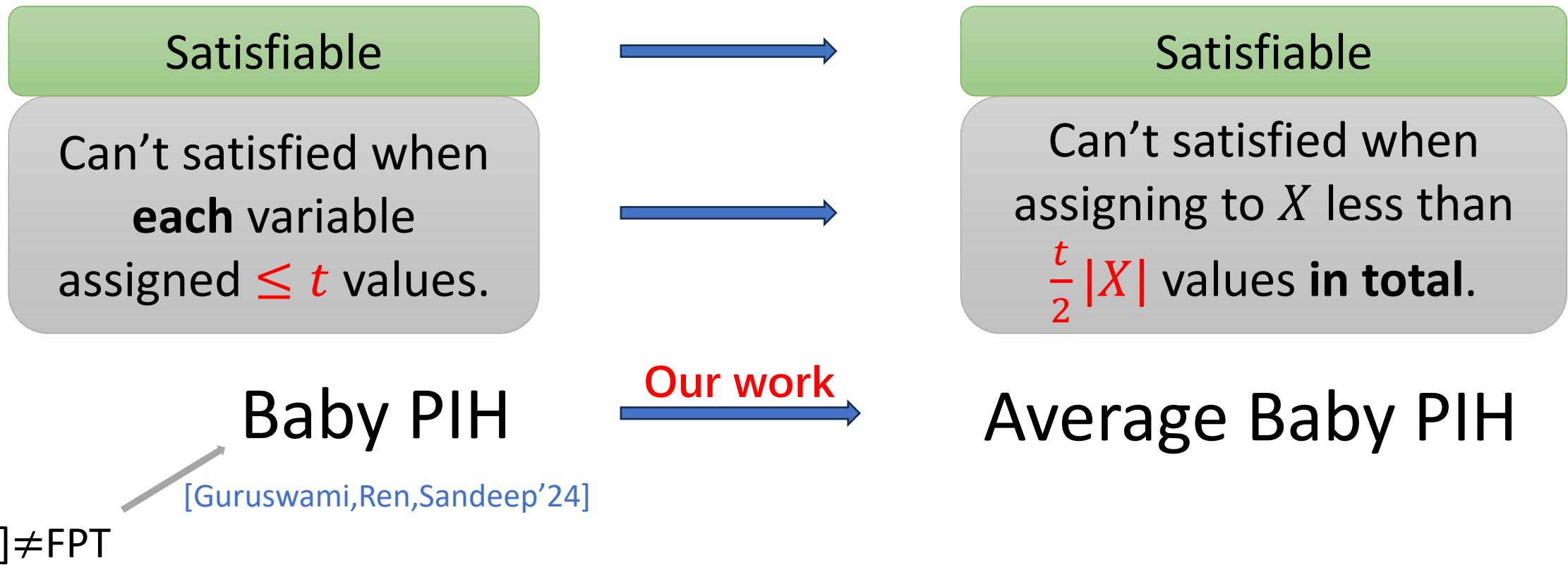
**Our work**

No FPT algorithm approximating **max num of  
assigned values to all** variable in 2CSP  
(Average Baby PIH)

[GRS24]

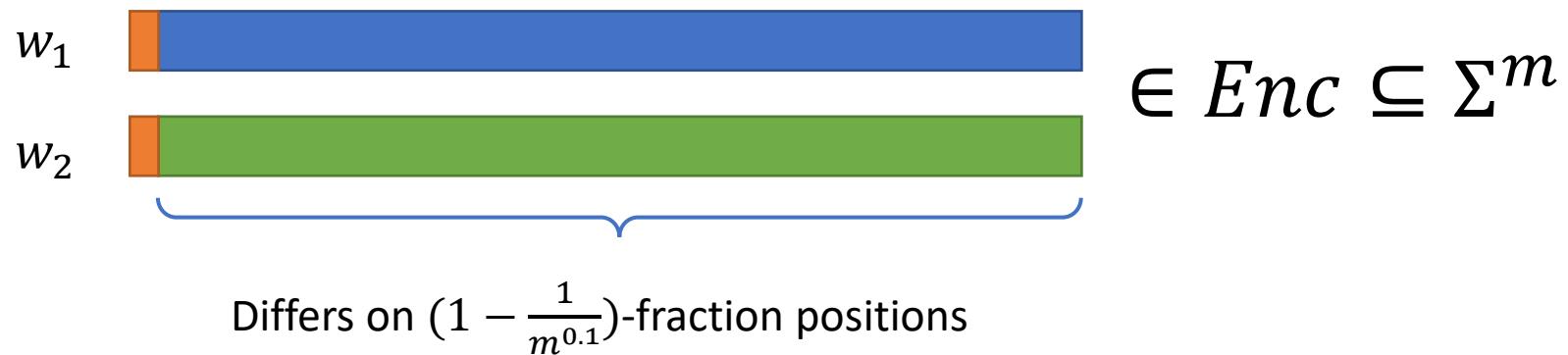
$\text{W}[1] \neq \text{FPT} \longrightarrow \text{Average Baby PIH}$

- A reduction for 2CSP instances that:



# Technical Tool

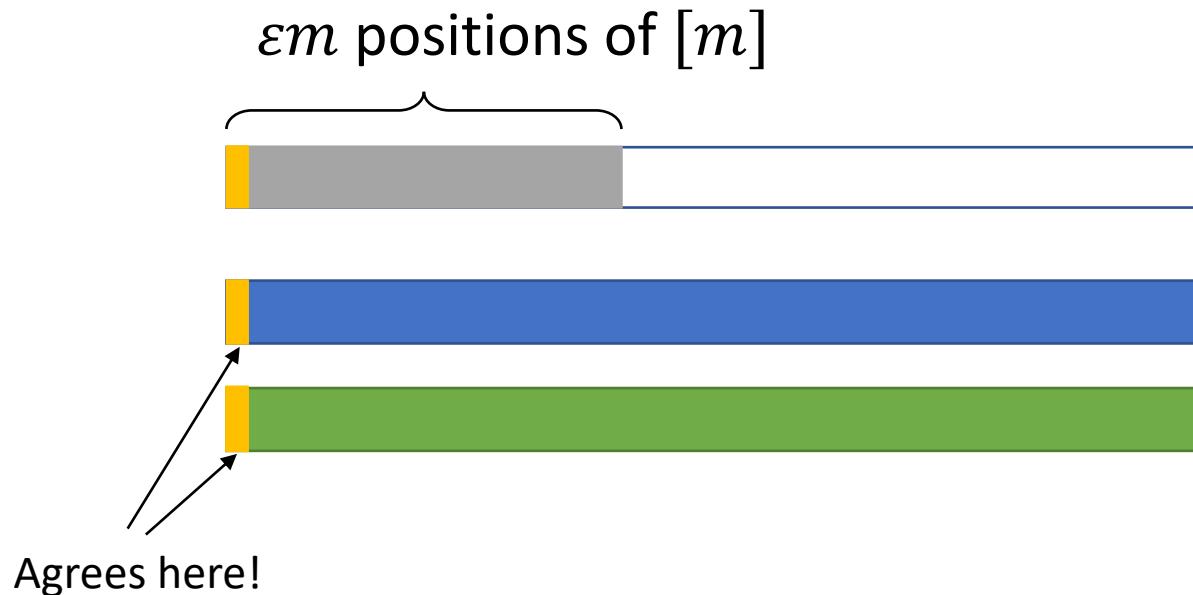
- Error-correcting codes with *overwhelming* (relative) distance



e.g. Reed-Solomon codes.

# Technical Tool

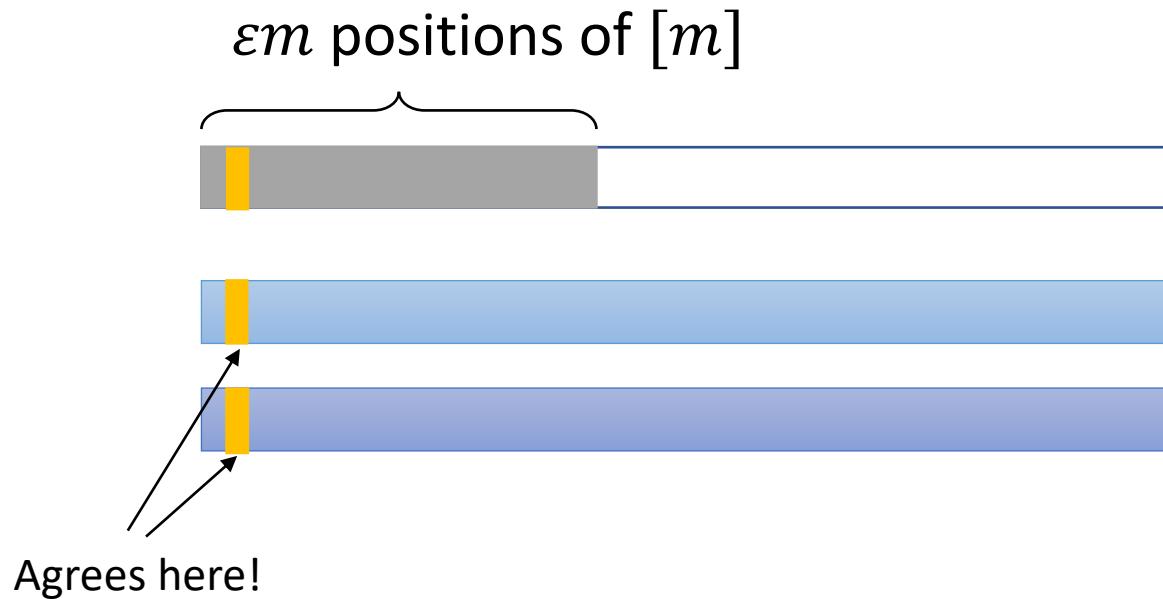
Any set  $S$  of codewords that “collides” on a noticeable fraction of positions.....



$$w_1 \in S \subseteq Enc \subseteq \Sigma^m$$
$$w_2$$

# Technical Tool

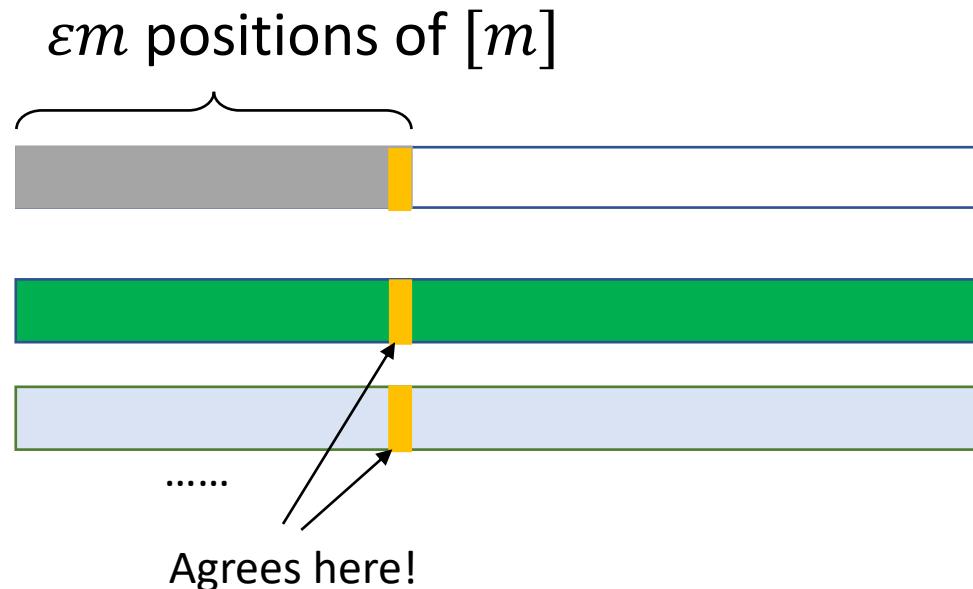
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$$w_3 \in S \subseteq Enc \subseteq \Sigma^m$$
$$w_4$$

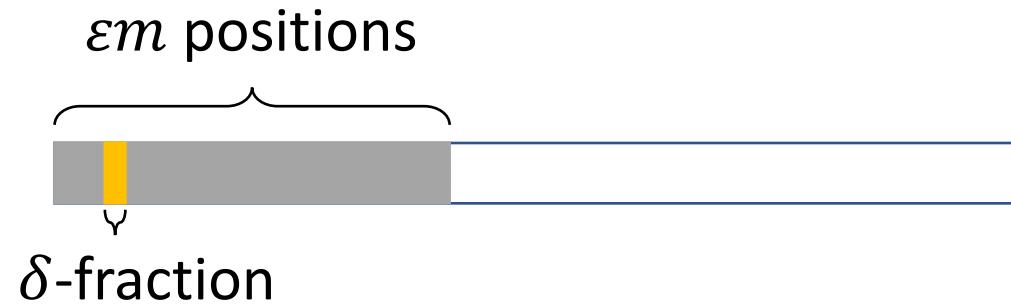
# Technical Tool

Any set  $S$  of codewords that “collides” on a noticeable fraction of positions.....



$$w_s \in S \subseteq Enc \subseteq \Sigma^m$$
$$w_{s+1}$$

# Technical Tool



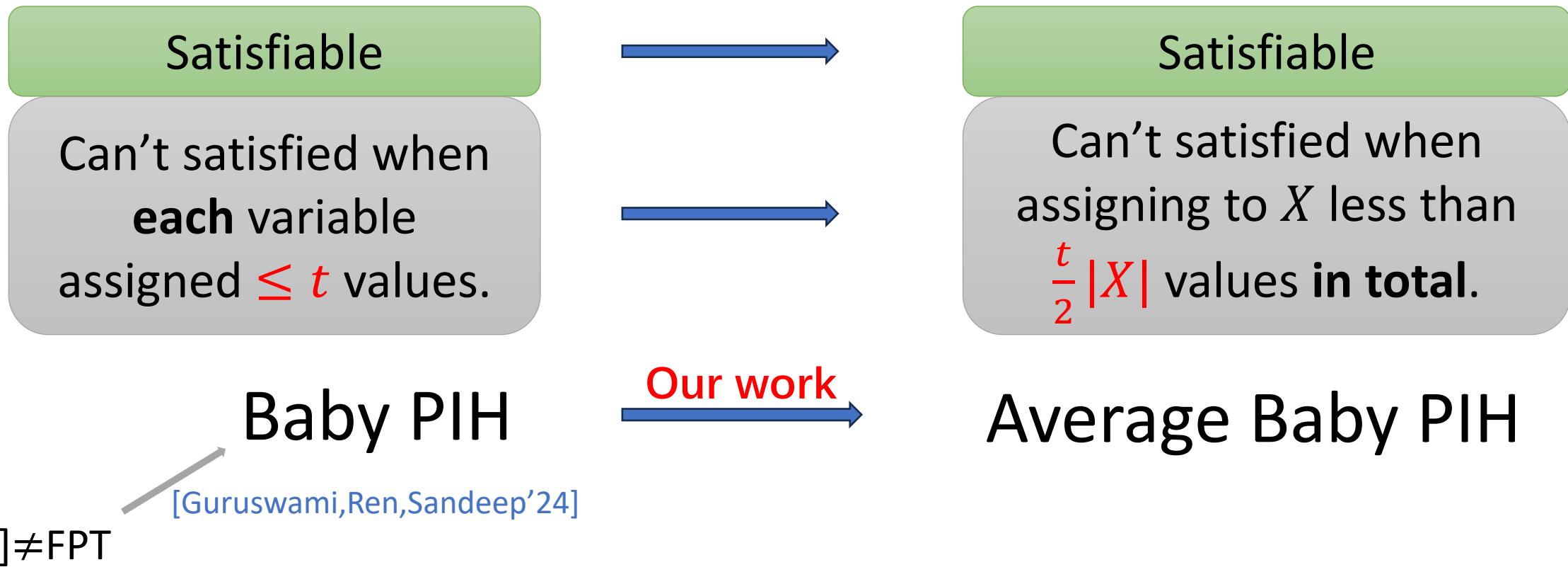
Theorem(Informal) cf. [Karthik-Navon'21, Lin-Ren-Sun-Wang'23]:

For code  $Enc$  with relative distance  $1 - \delta$ , any set of codewords “collides”

on  $\varepsilon m$  positions must have size  $\geq \sqrt{\frac{2\varepsilon}{\delta}}$ .

Recall:  $\text{W}[1] \neq \text{FPT} \longrightarrow$  Average Baby PIH

- A reduction for 2CSP instances that:

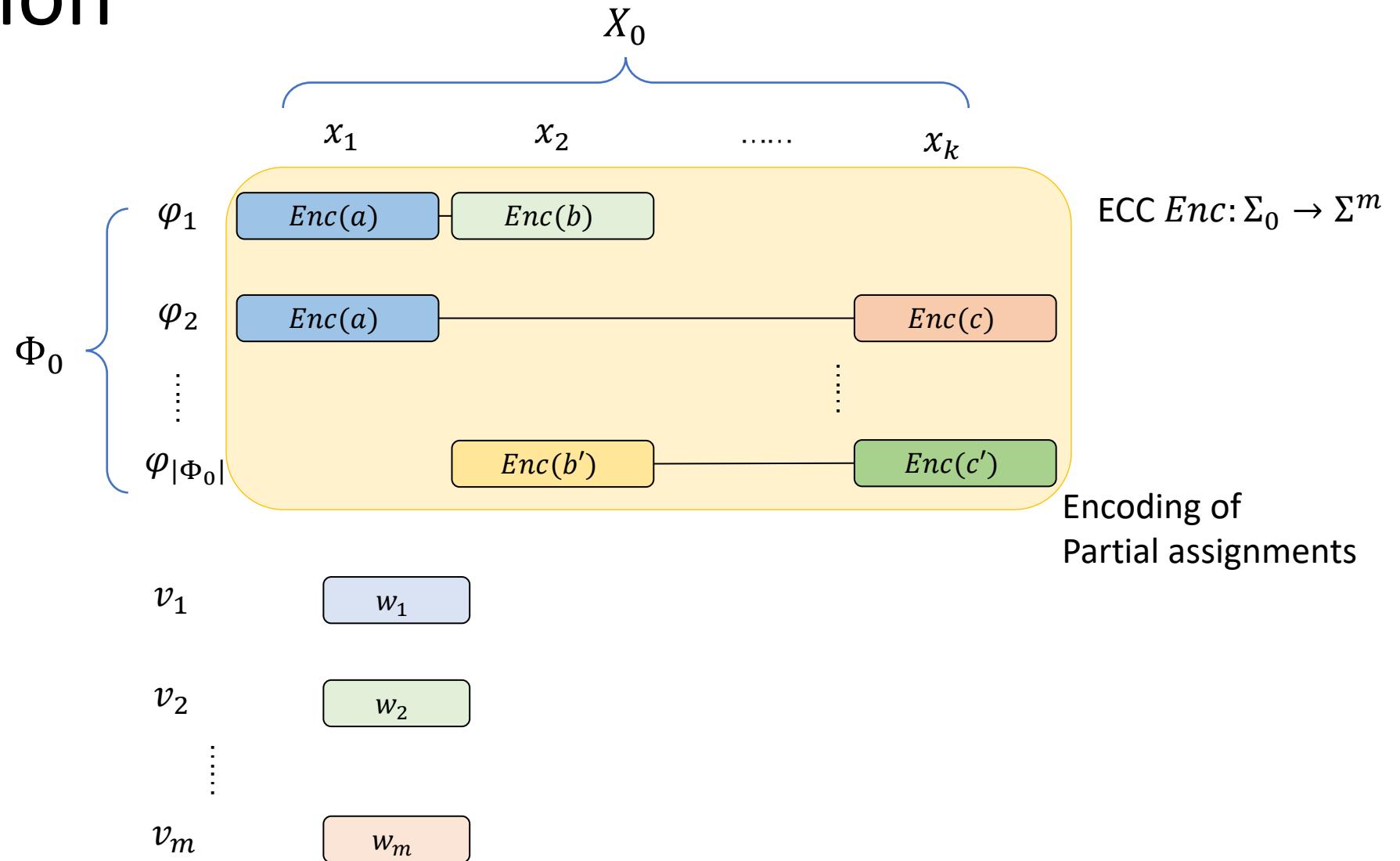


# The Reduction

Input: 2CSP instance  
 $\Pi_0 = (X_0, \Sigma_0, \Phi_0)$

Output: 2CSP instance  $\Pi$   
as shown.

**Variables:**  $\Phi_0 \cup \{v_1, \dots, v_m\}$



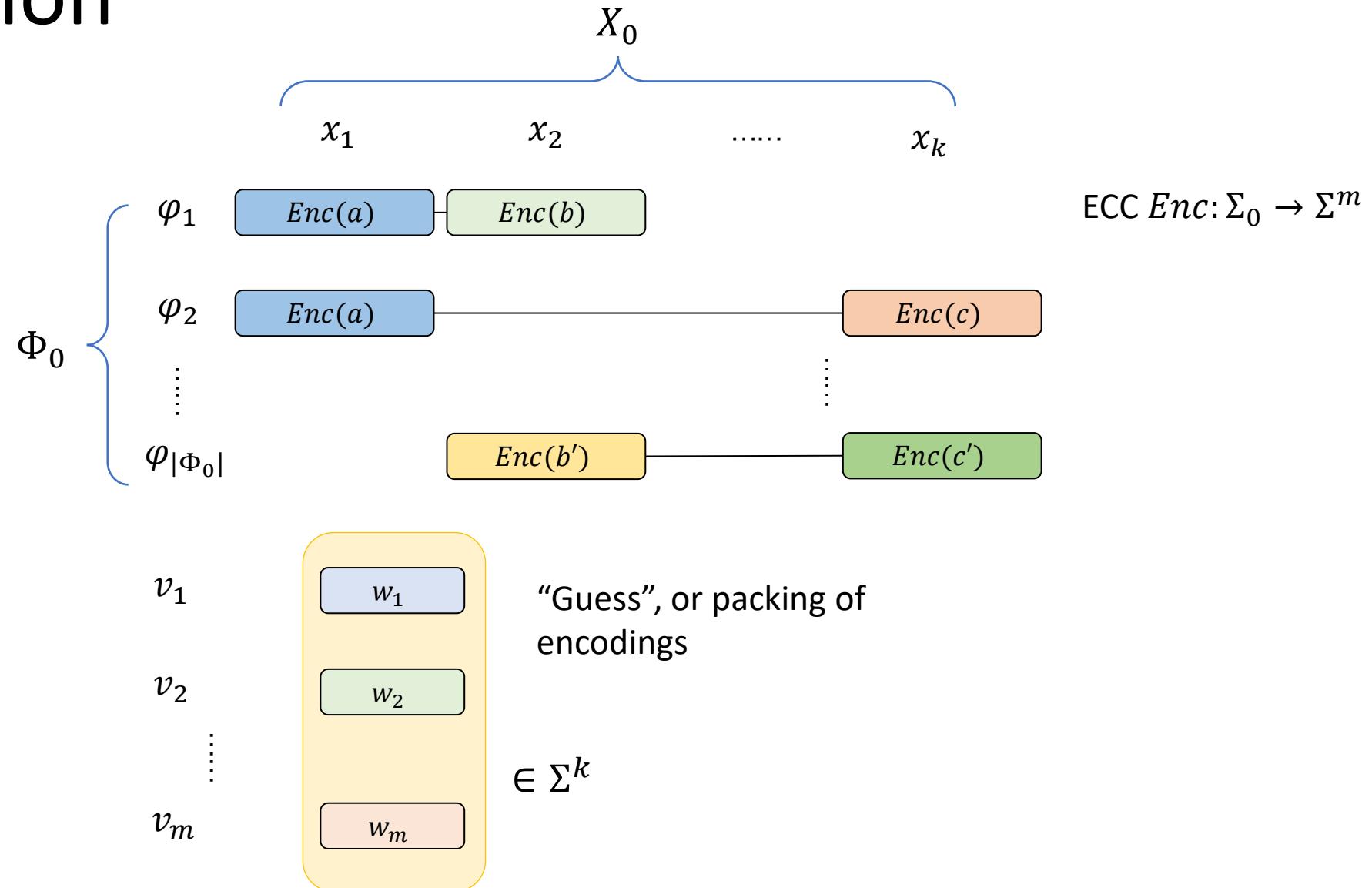
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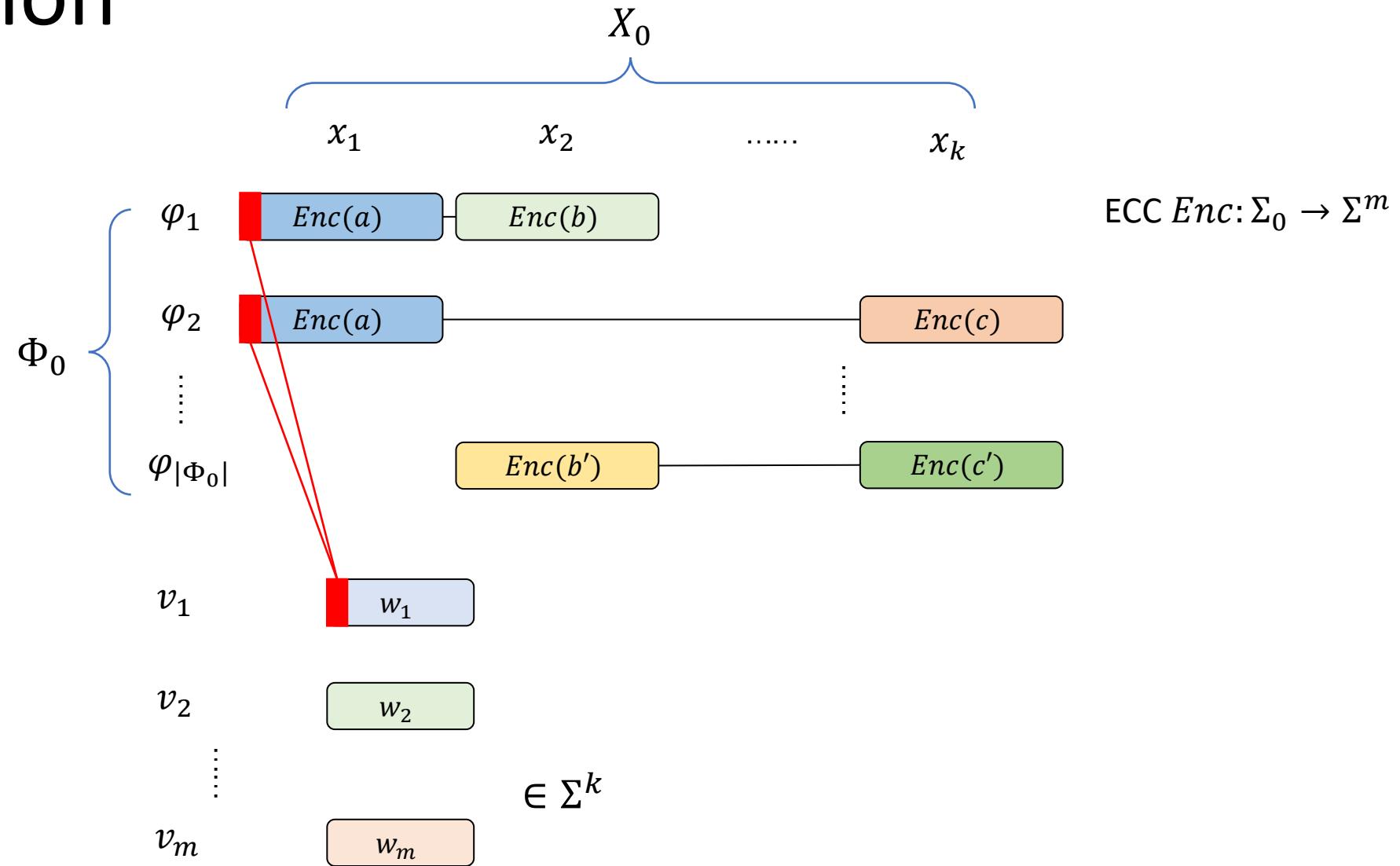


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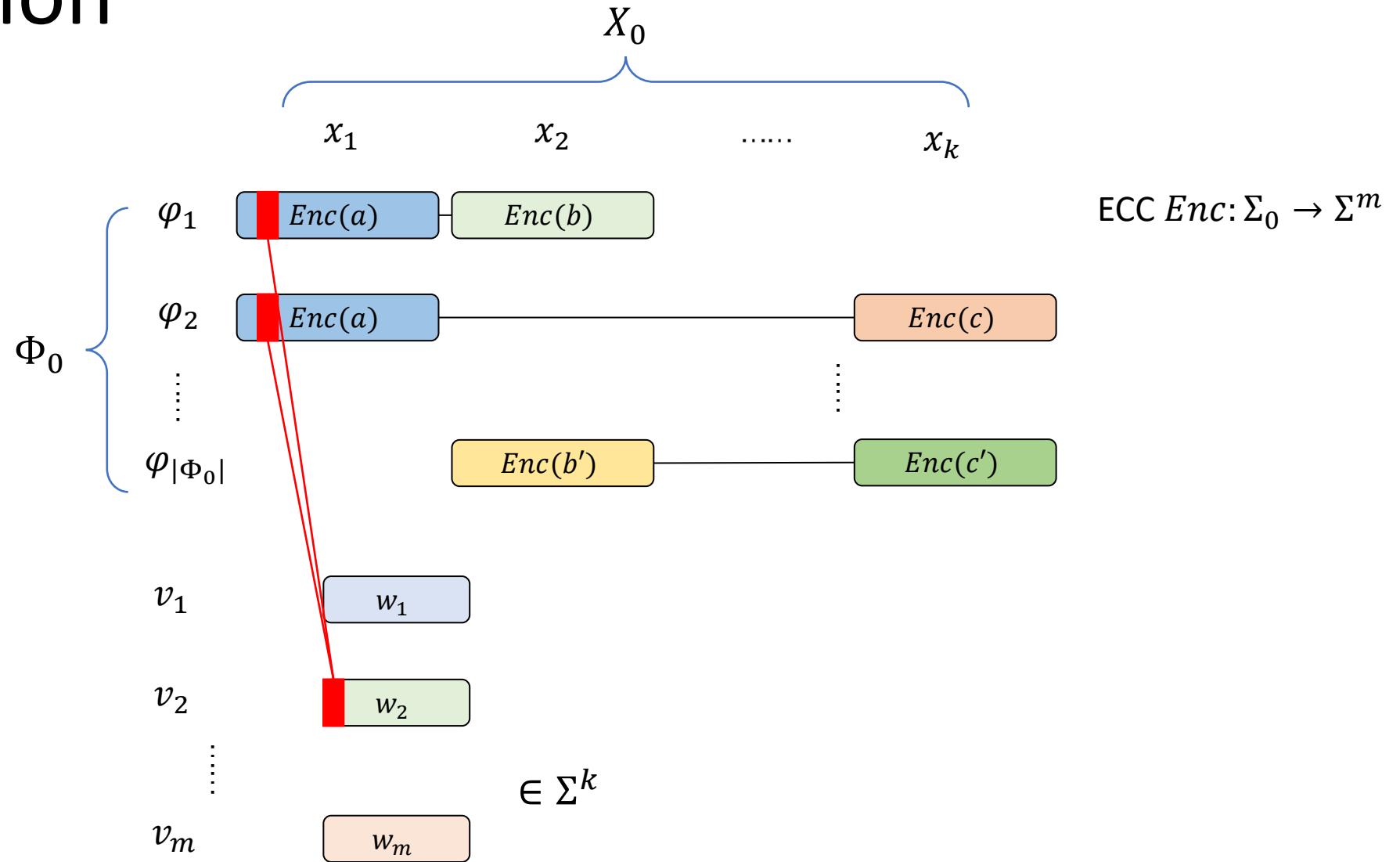
**Constraints: Equality Check**

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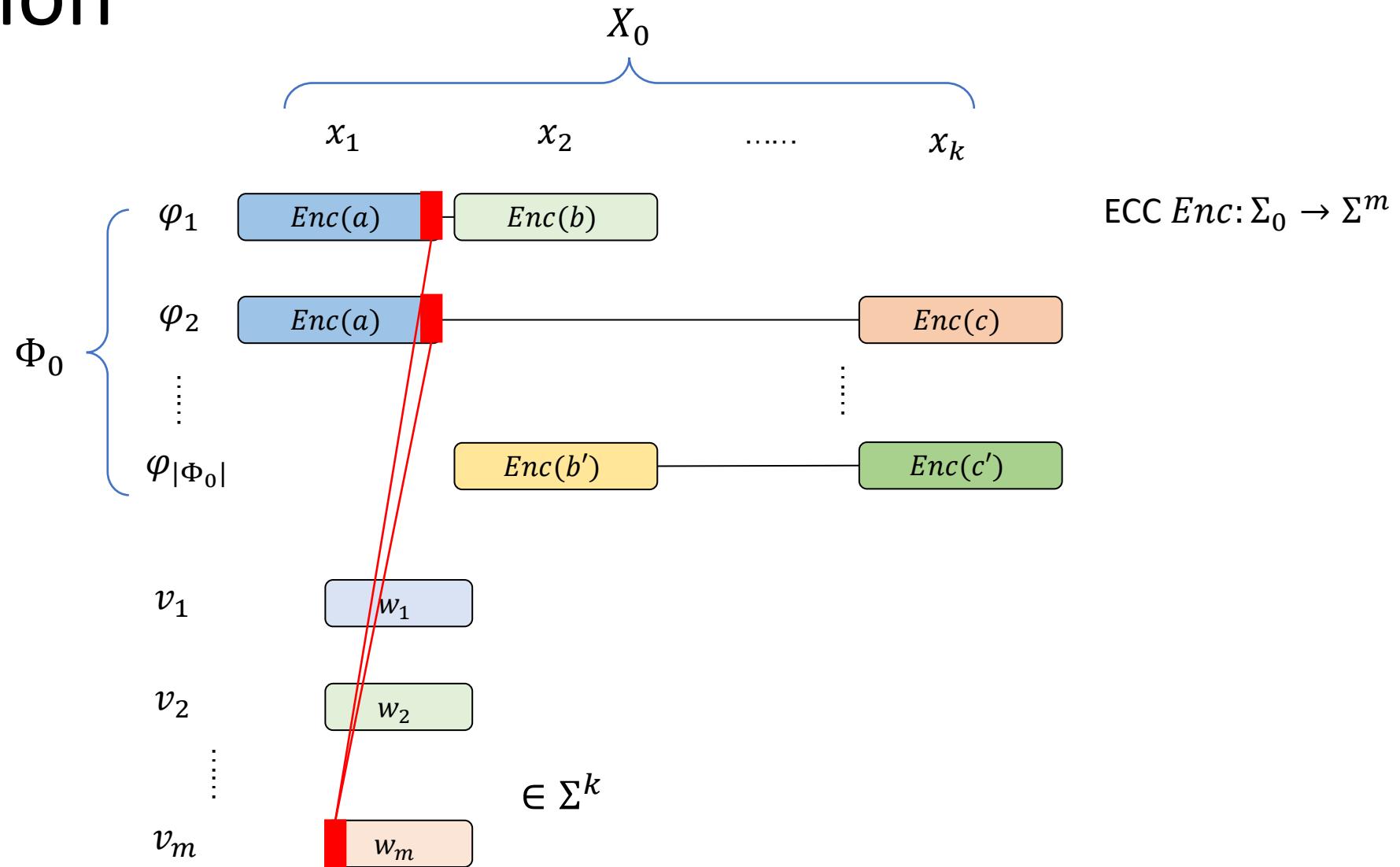
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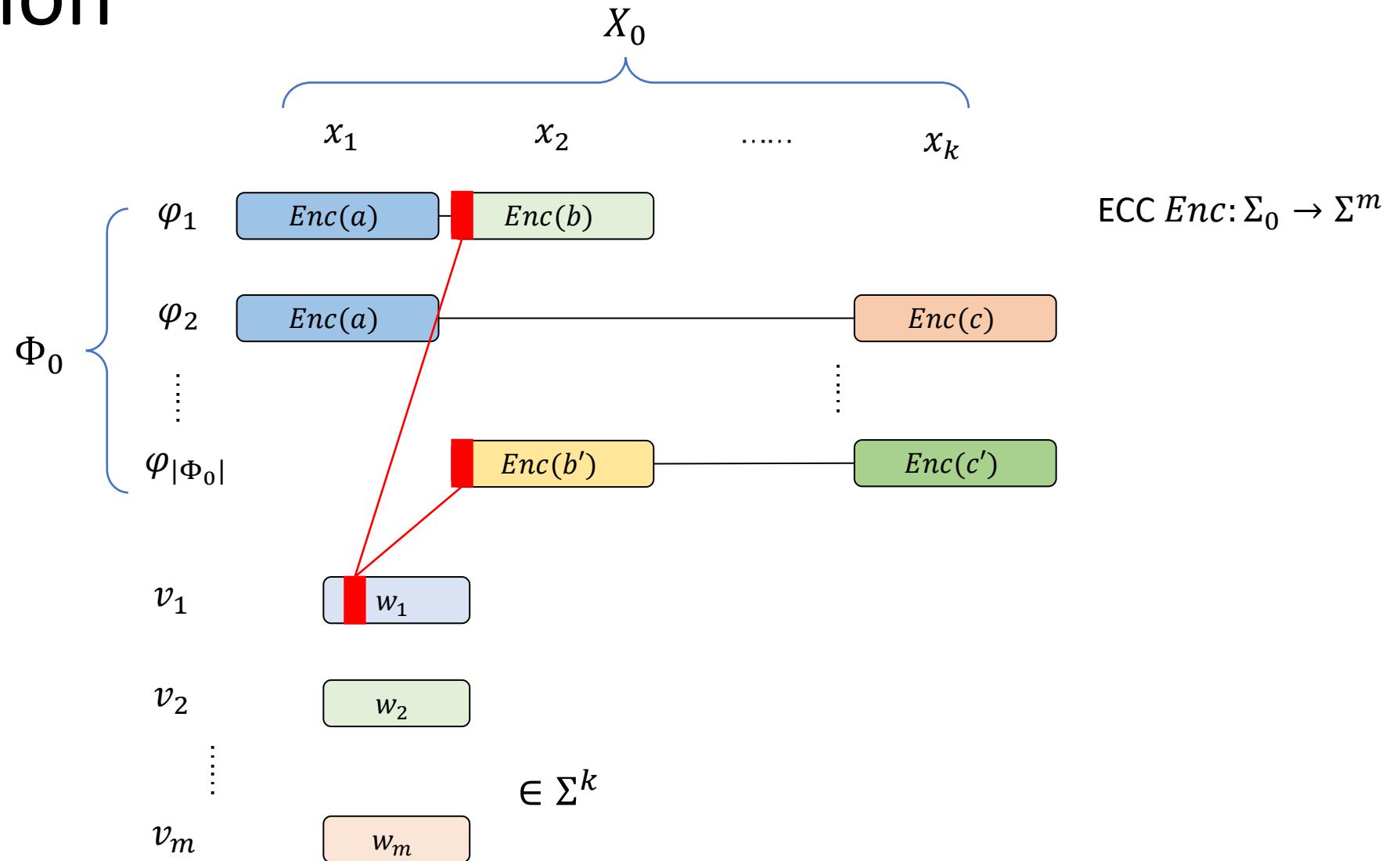


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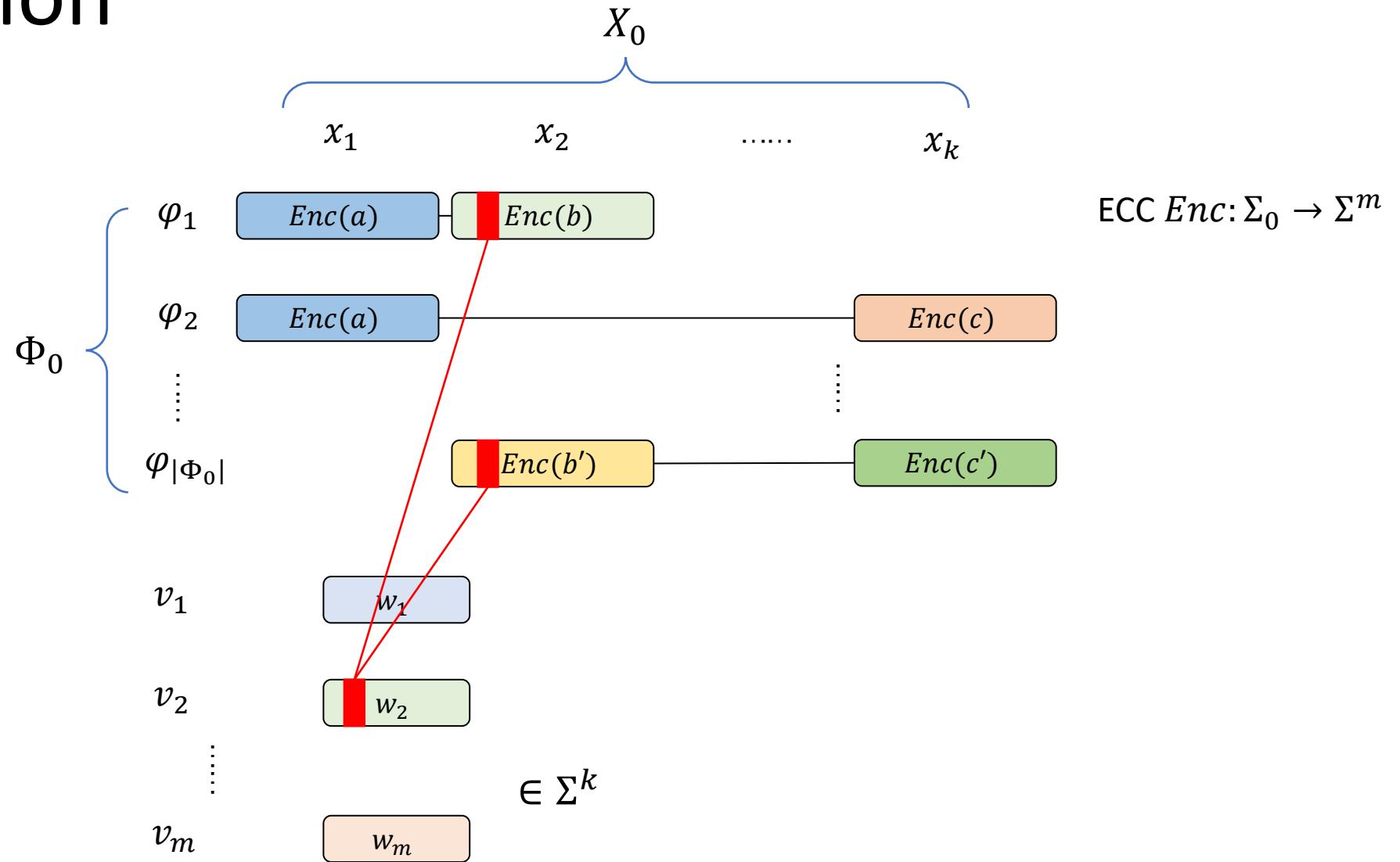


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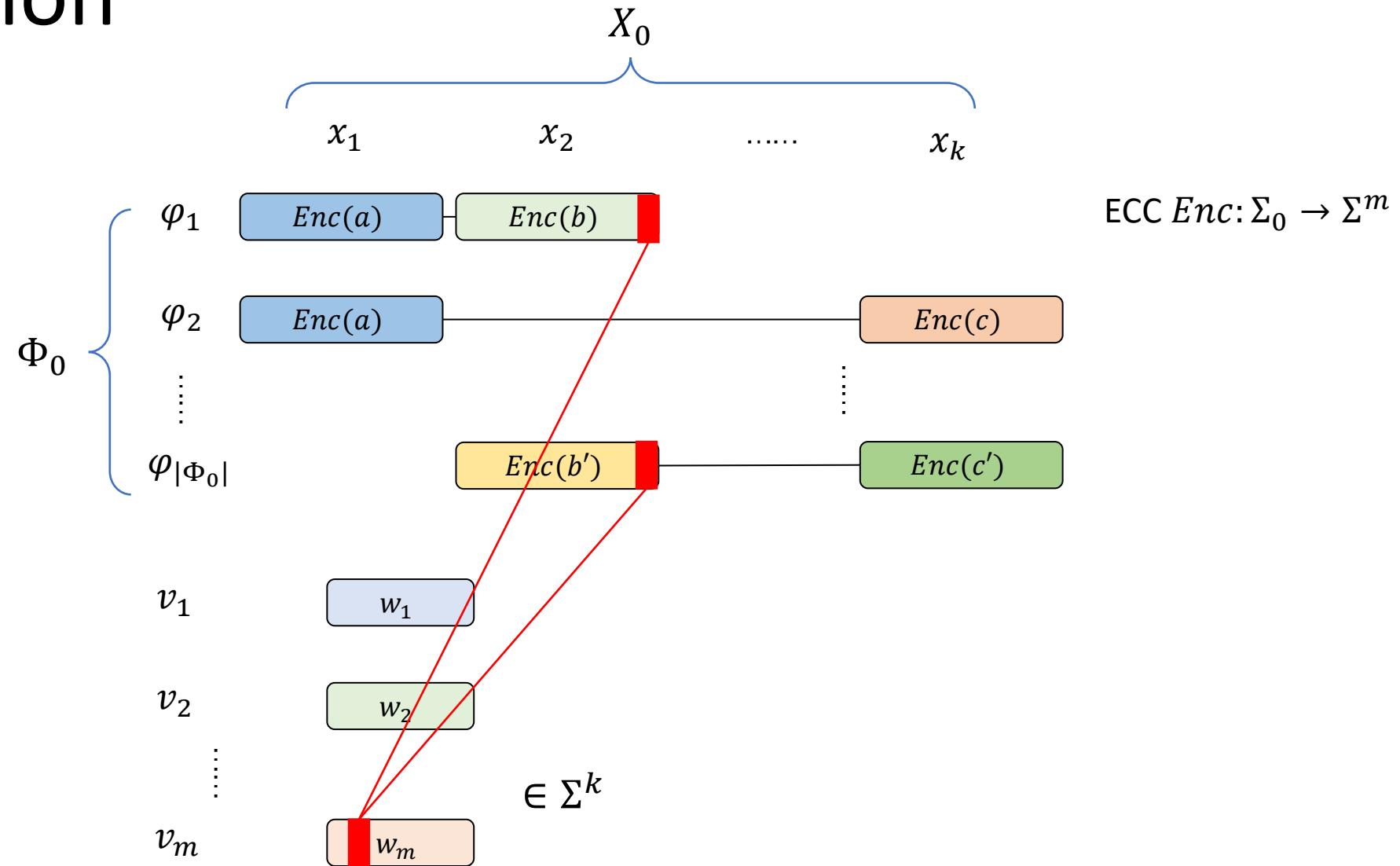
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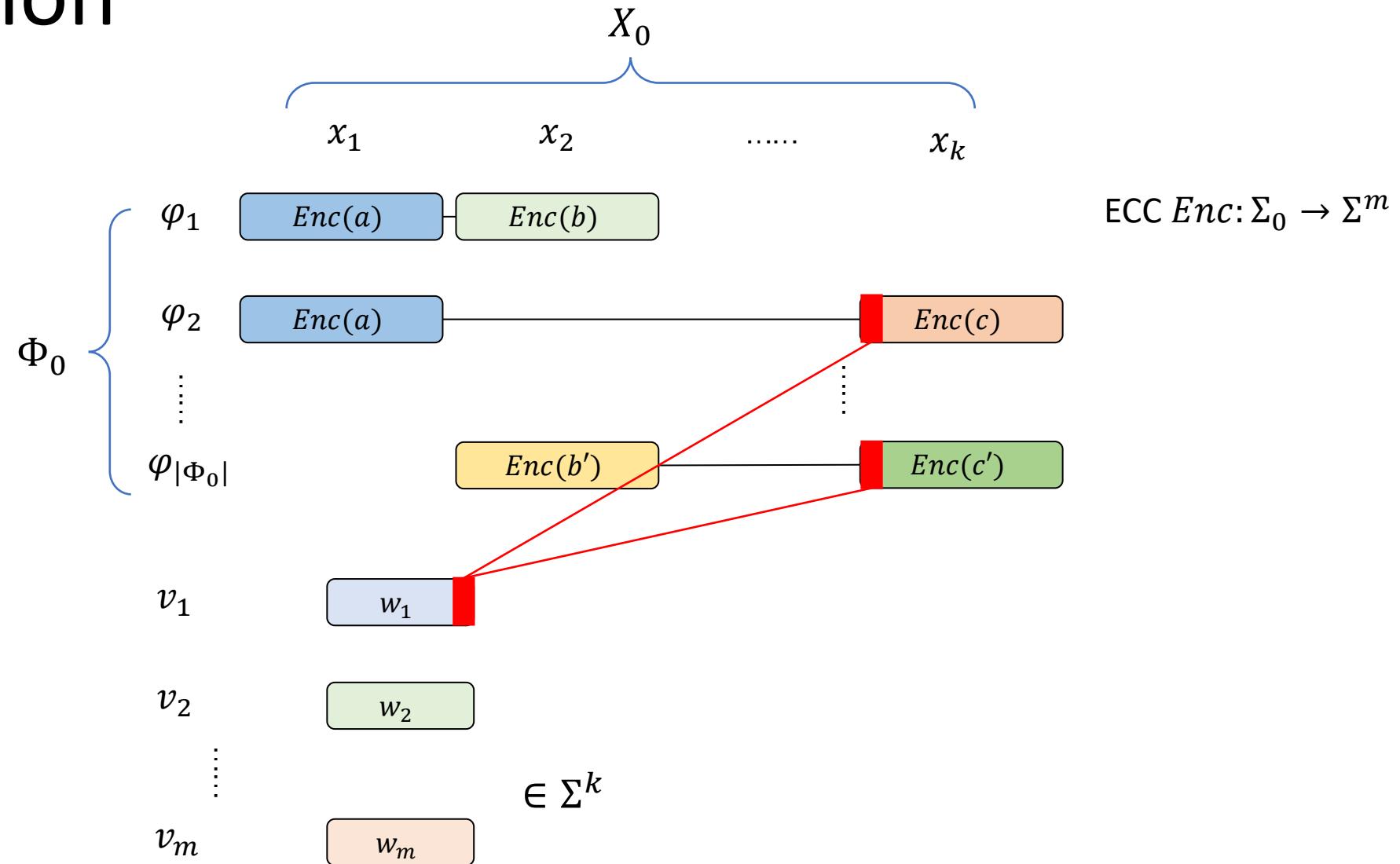
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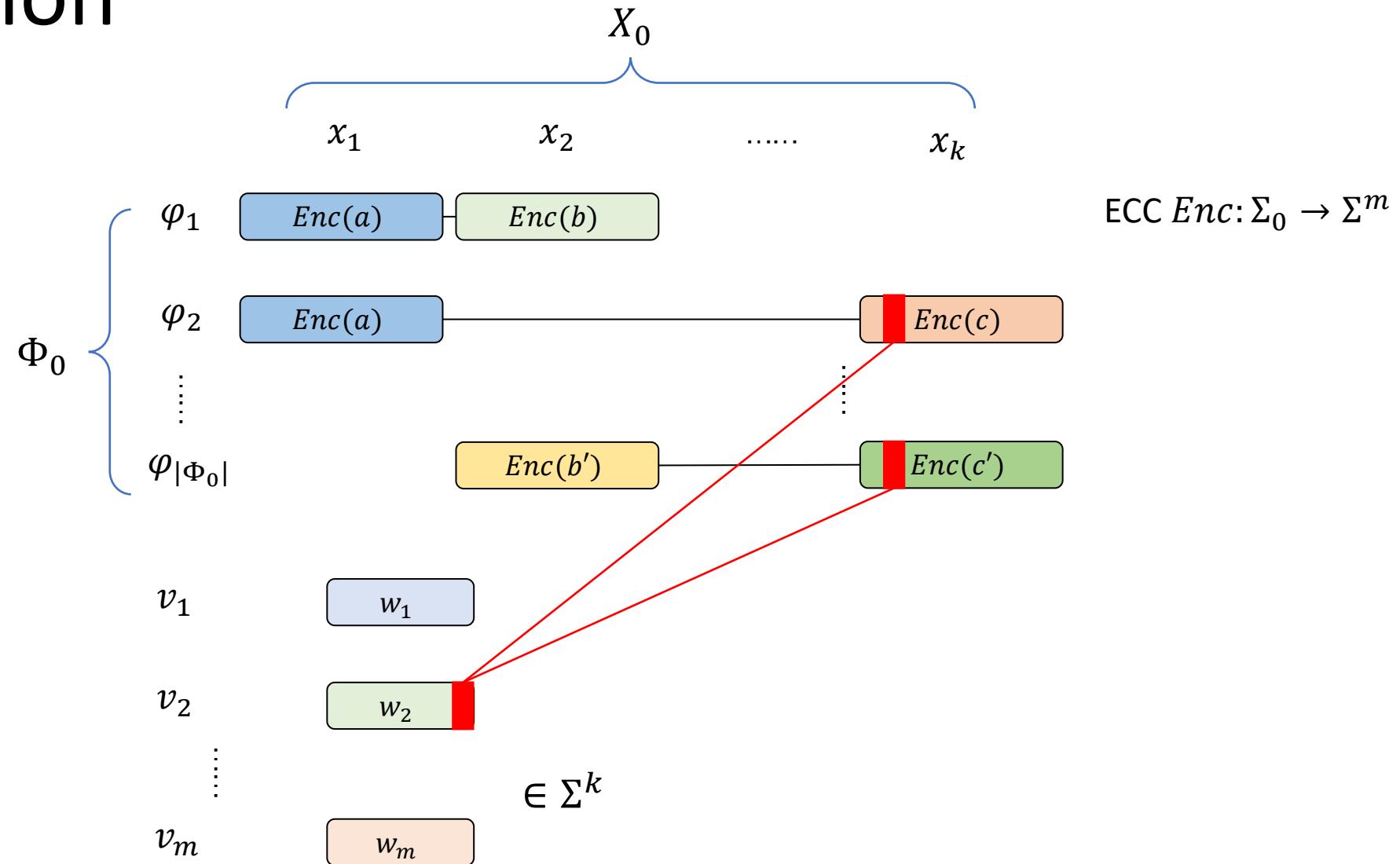
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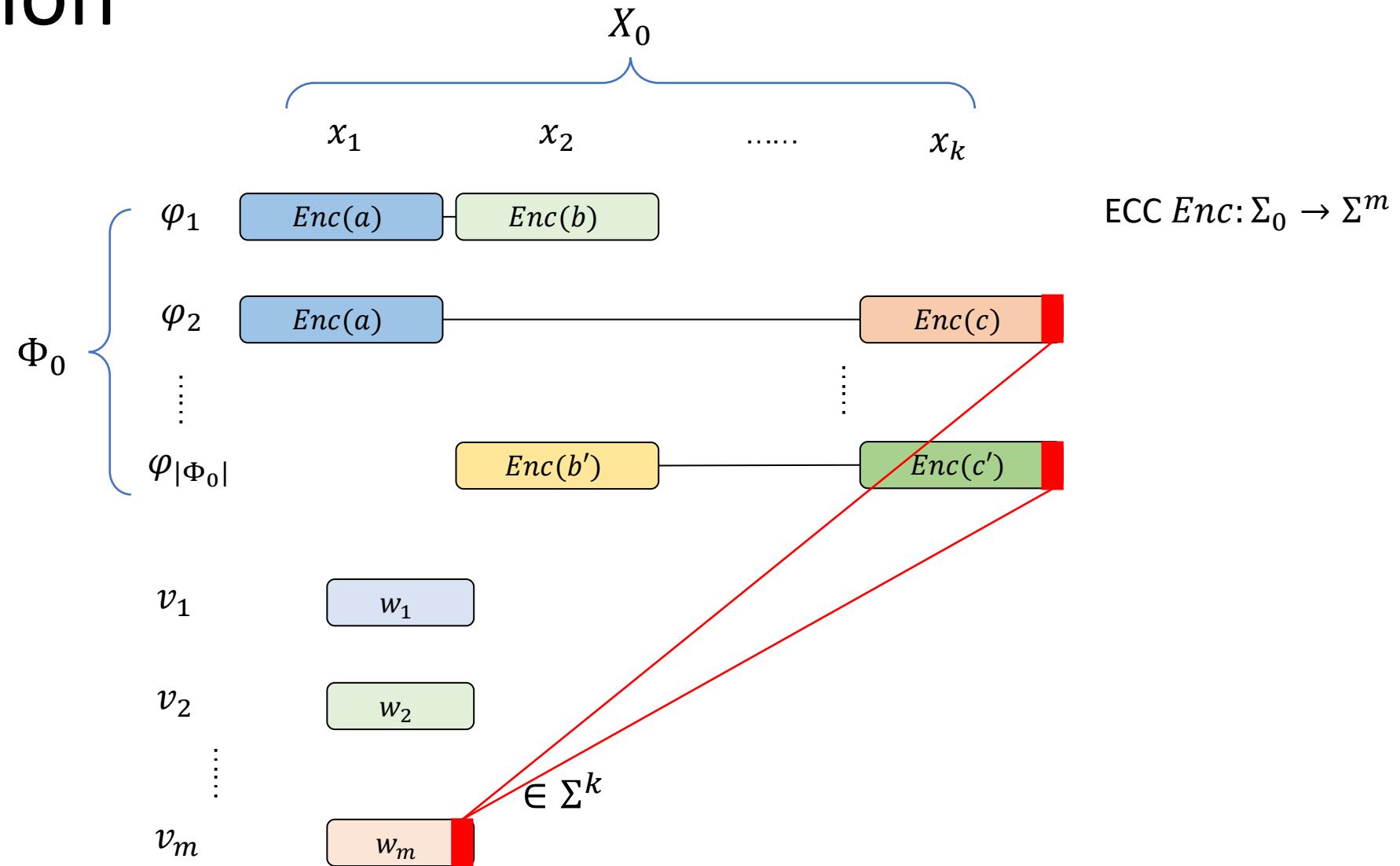
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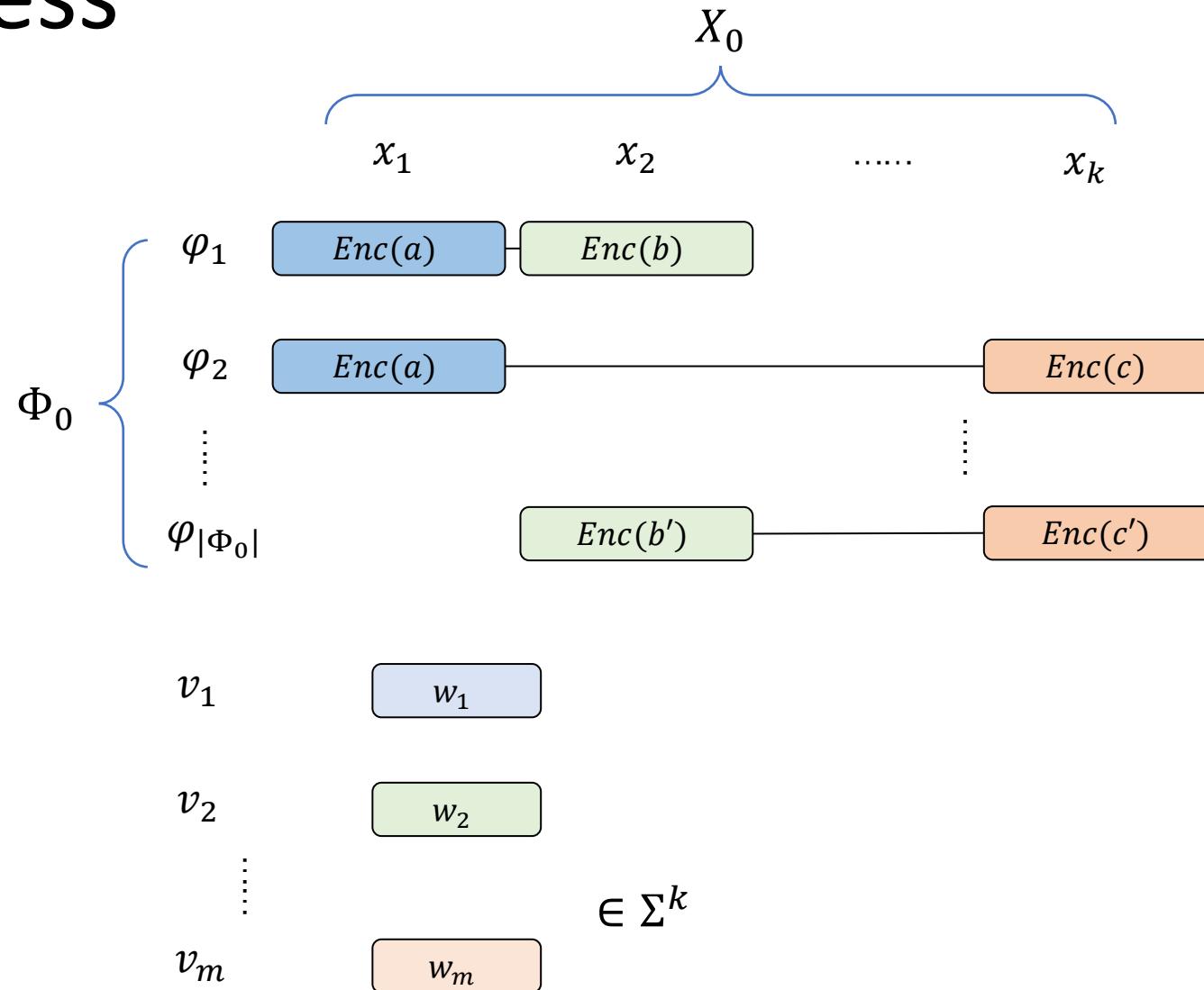


# Completeness

$\Pi_0 = (X_0, \Sigma_0, \Phi_0)$   
Satisfiable

Direct

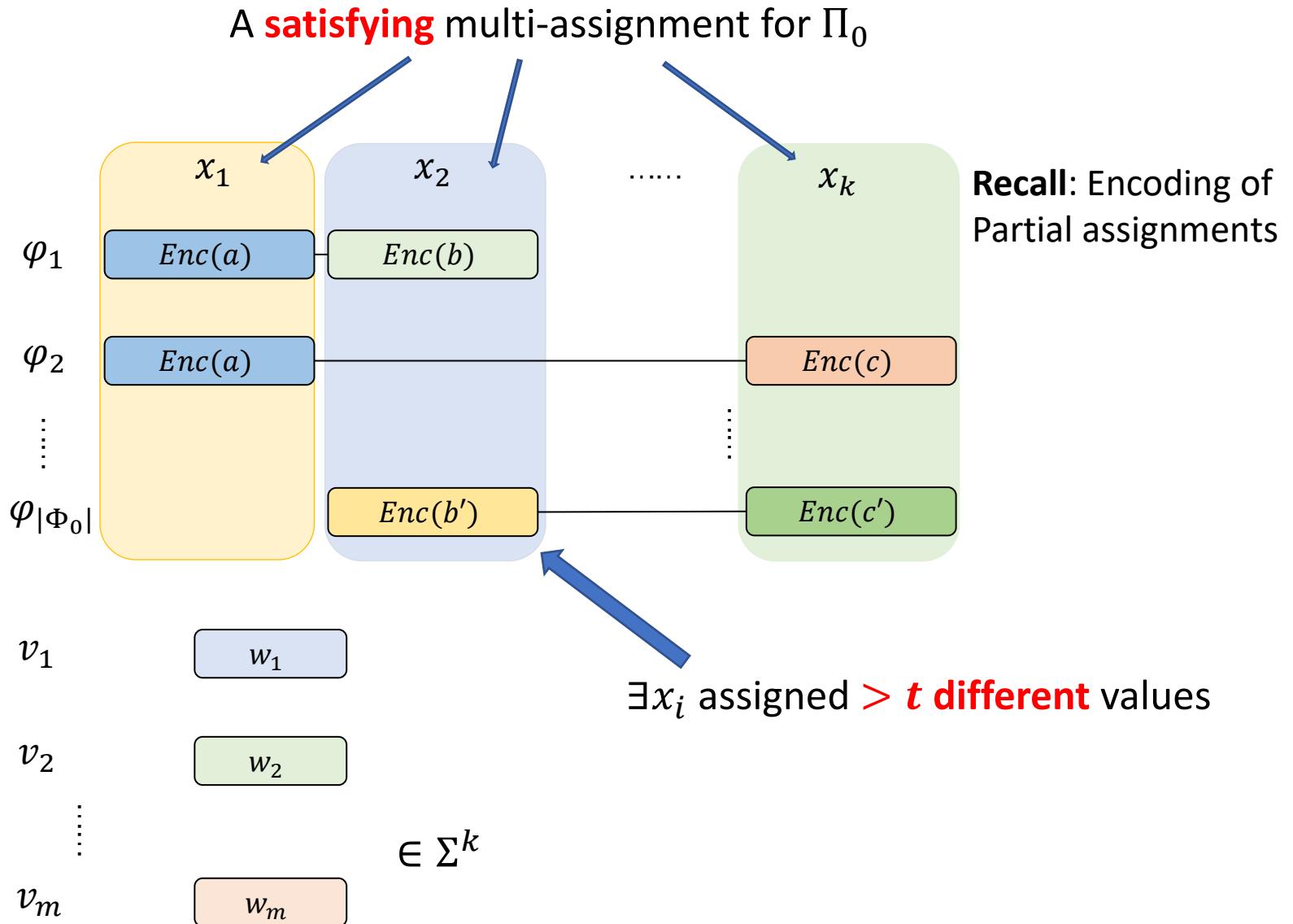
$\Pi$  Satisfiable



# Soundness

$$\Pi_0 = (X_0, \Sigma_0, \Phi_0)$$

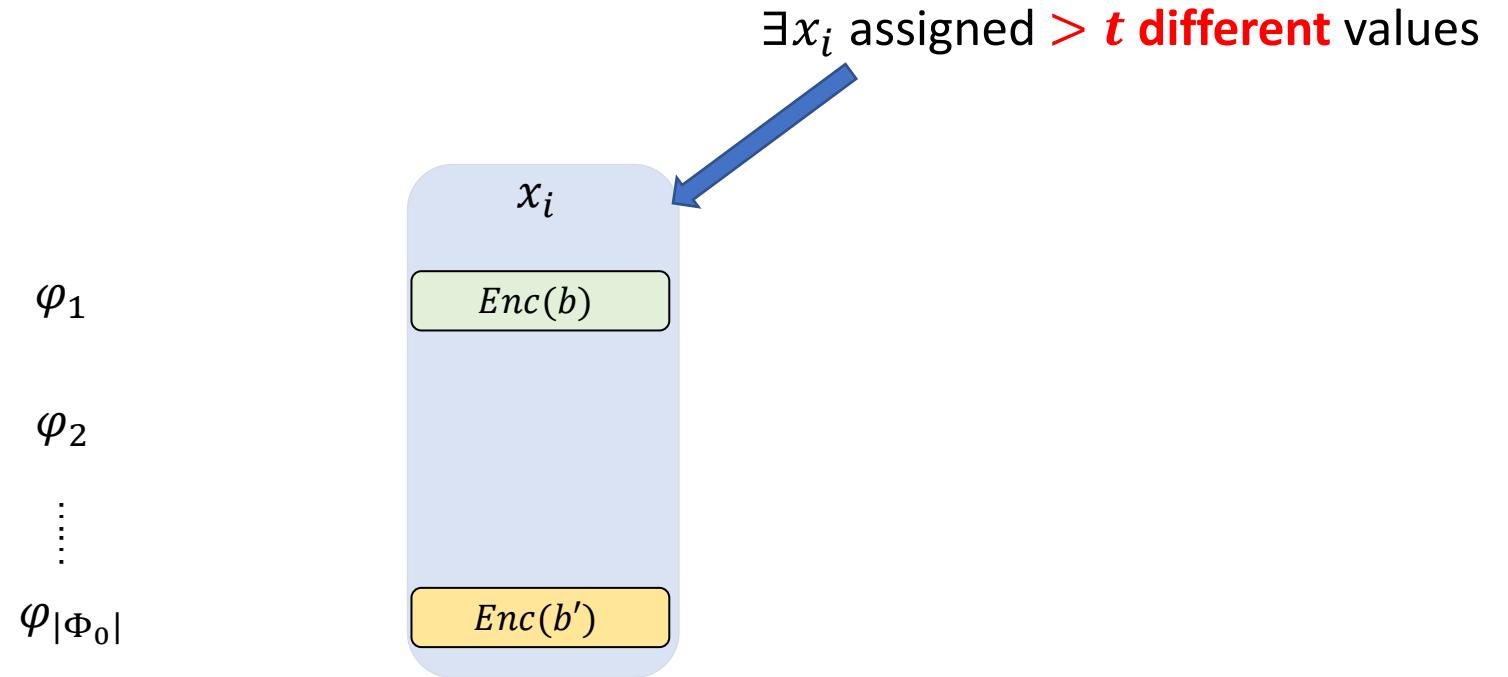
Can't satisfied when **each** variable assigned  $\leq t$  values



# Soundness

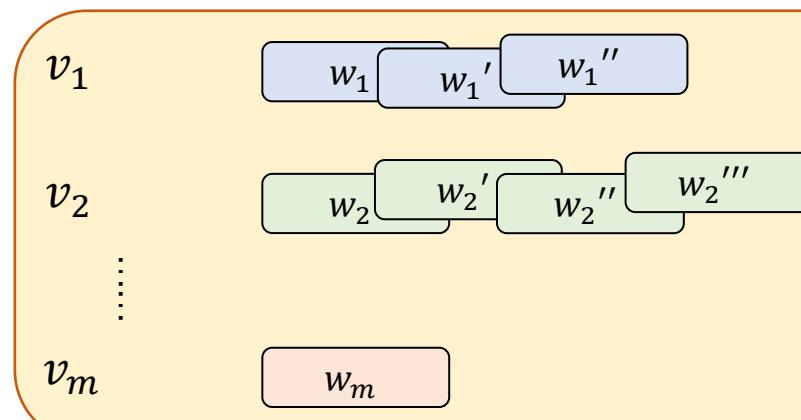
$$\Pi_0 = (X_0, \Sigma_0, \Phi_0)$$

Can't satisfied when **each** variable assigned  $\leq t$  values



Case 1:

**More than  $(1 - \varepsilon)$  fraction** of  $v$ 's, each assigned  $t + 1$  values



**Total # of values:**  
 $\geq (1 - \varepsilon)t \cdot m$

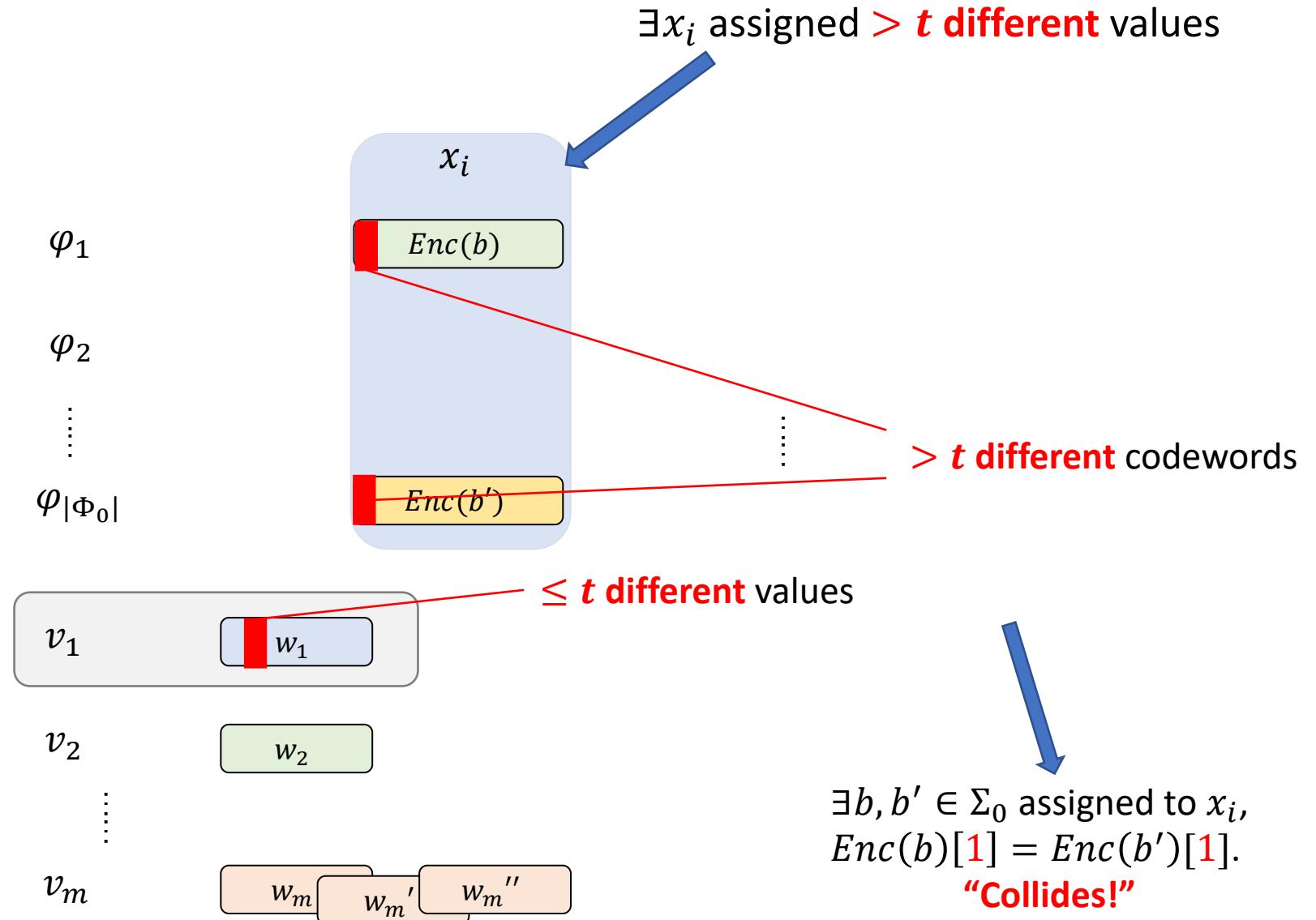
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Can't satisfied when **each** variable assigned  $\leq t$  values

Case 2:

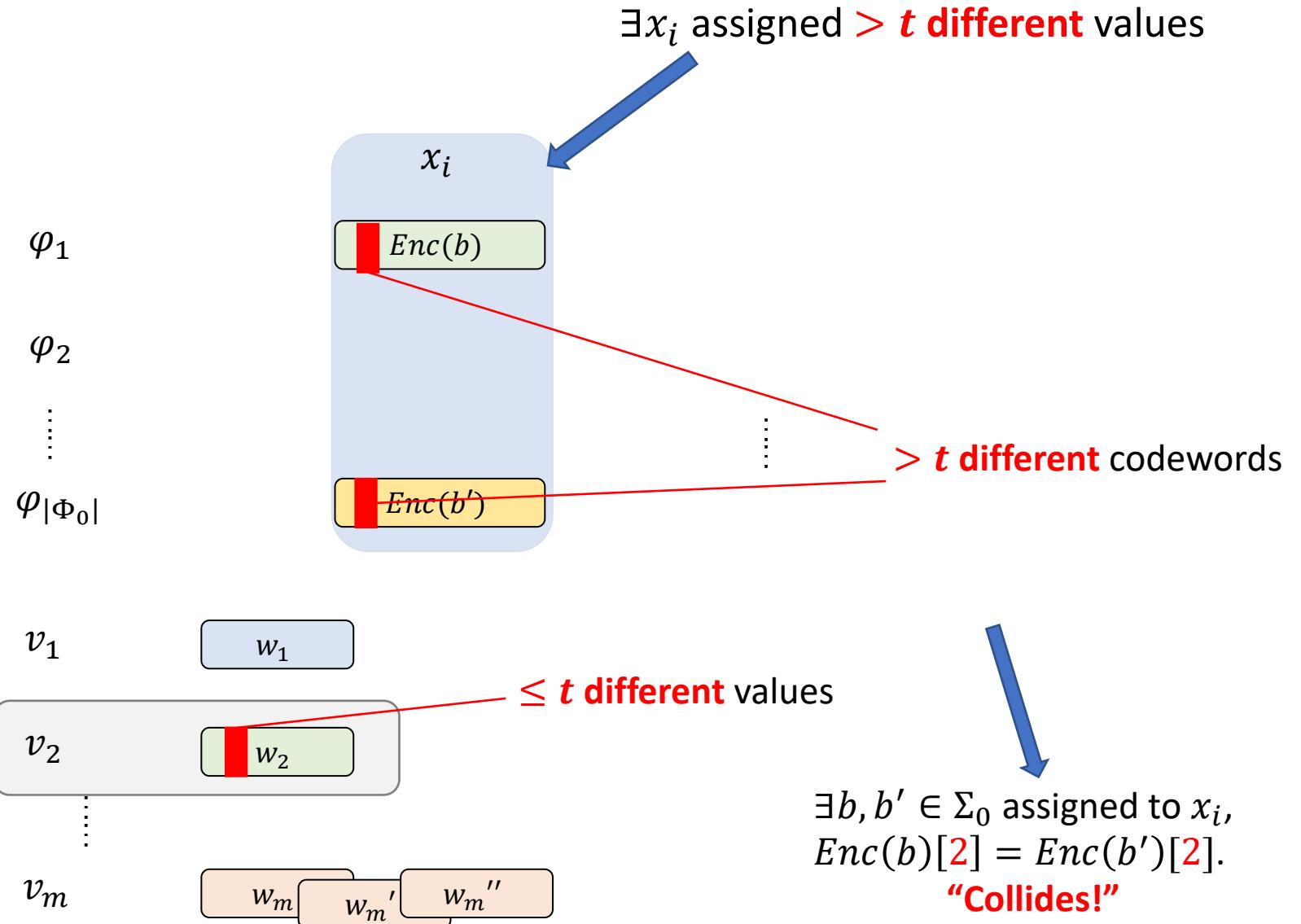
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# Soundness

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Case 2:

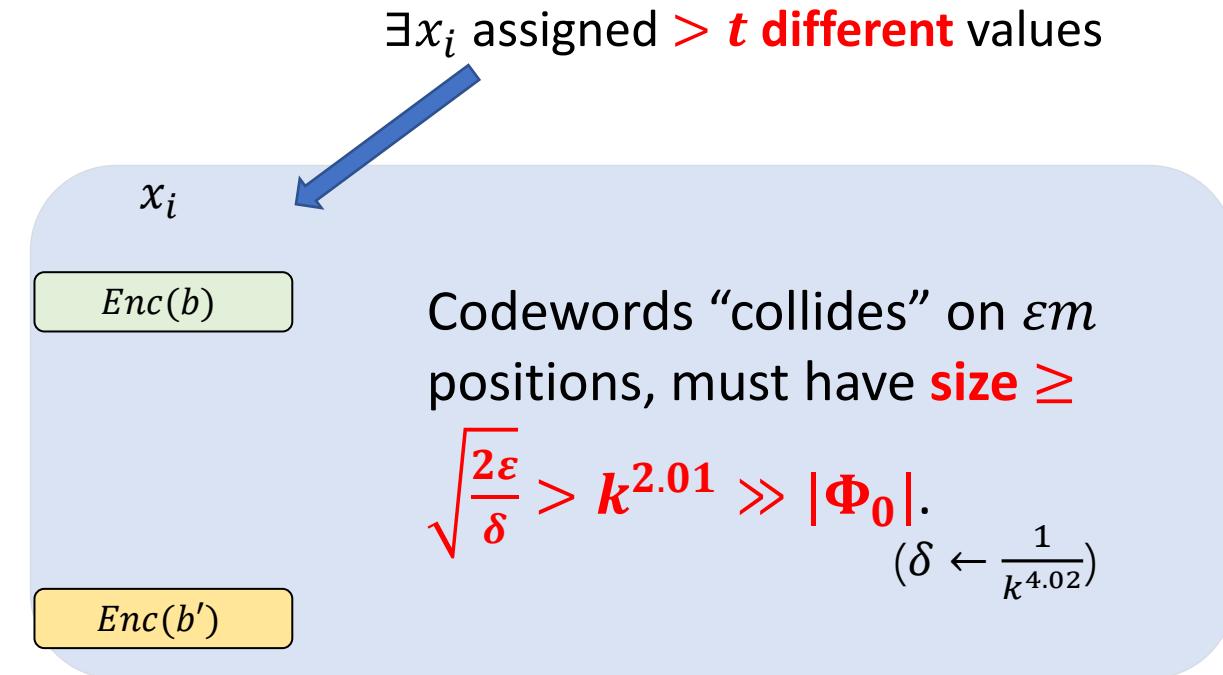
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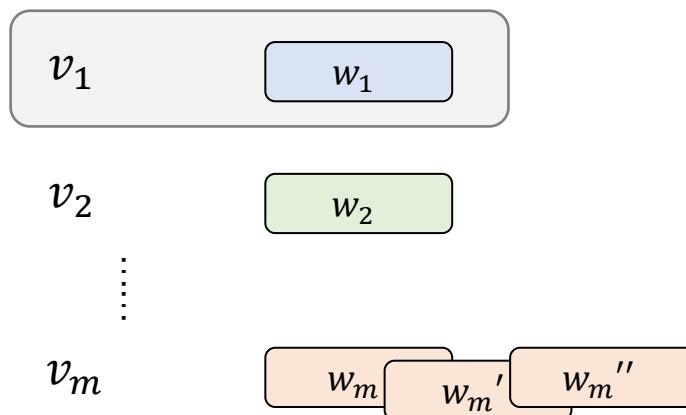
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$\varphi_1$   
 $\varphi_2$   
 $\vdots$   
 $\varphi_{|\Phi_0|}$



Case 2:

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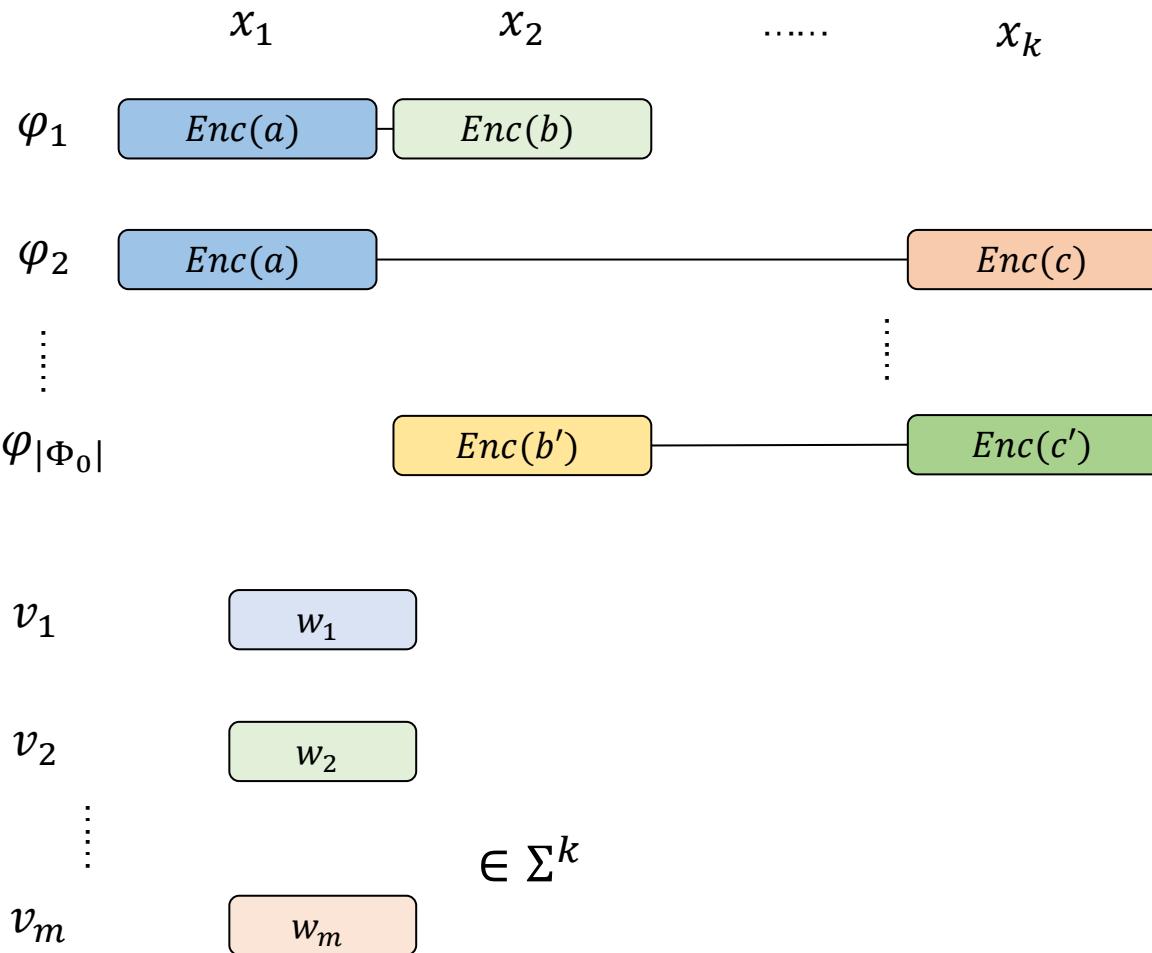
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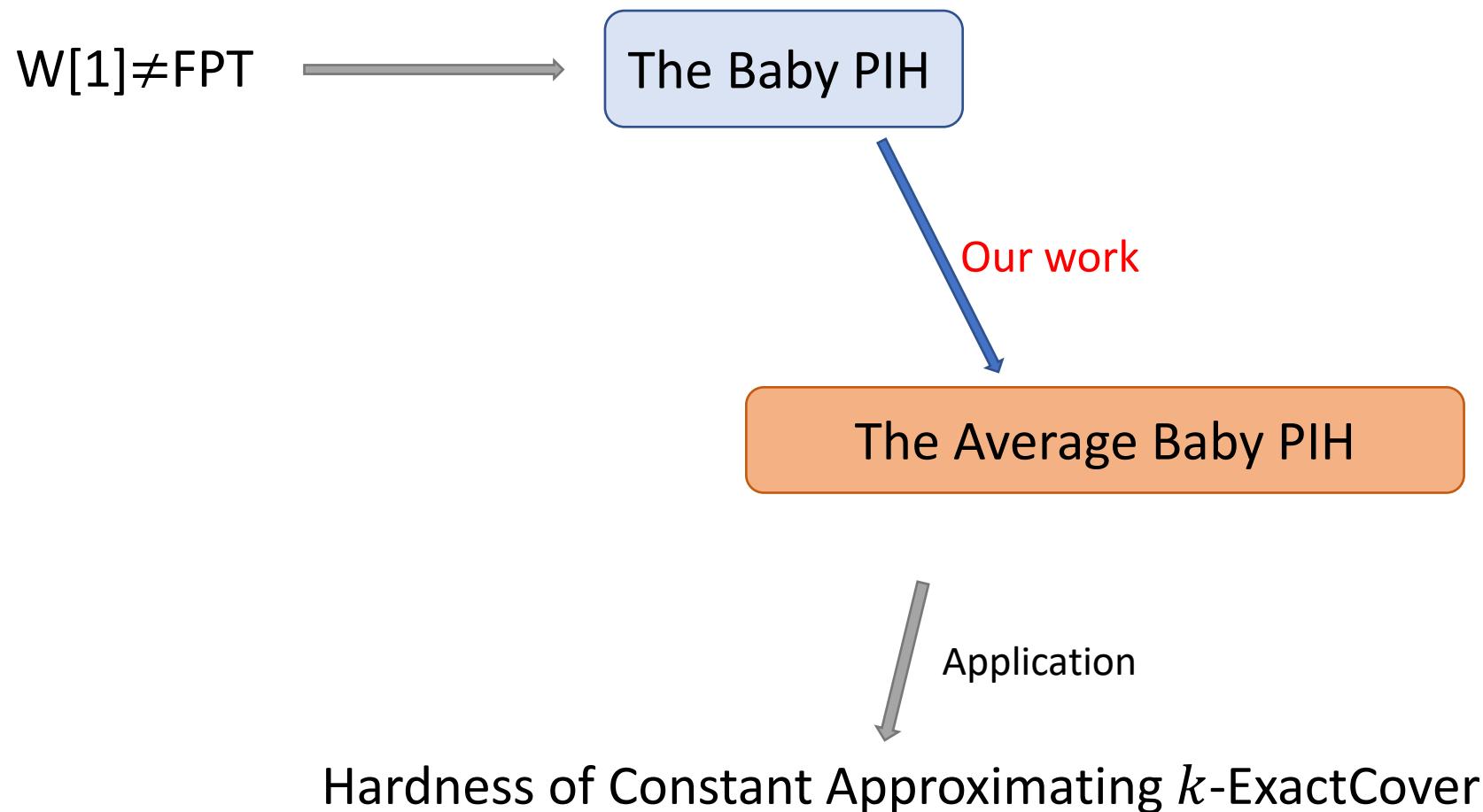
Can't satisfied when **each** variable assigned  $\leq t$  values



$\Pi$  Can't satisfied when assigning to  $X$  less than  $\min(\frac{t}{2}|X|, k^2)$  values **in total**.



# Conclusion



# Open Question

$W[1] \neq FPT$

Our work

The Average Baby PIH  
For  $\Pi = (X, \Sigma, \Phi)$  with  
 $|\Phi| = \omega(|X|)$

(Pointed out by reviewers)

The Average Baby PIH  
For  $\Pi = (X, \Sigma, \Phi)$  with  
 $|\Phi| = O(|X|)$

Implies

**The PIH**

**Thank You!**