

Ilaria Zappatore

Simultaneous Rational Function Reconstruction and applications to Algebraic Coding Theory

a joint work with

Eleonora GUERRINI, Romain LEBRETON LIRMM, Université de Montpellier CNRS

GT de l'équipe GRACE

December 1, 2020

Starting Point

Coding
Theory

My
Thesis



Computer
Algebra

Overview & Motivations

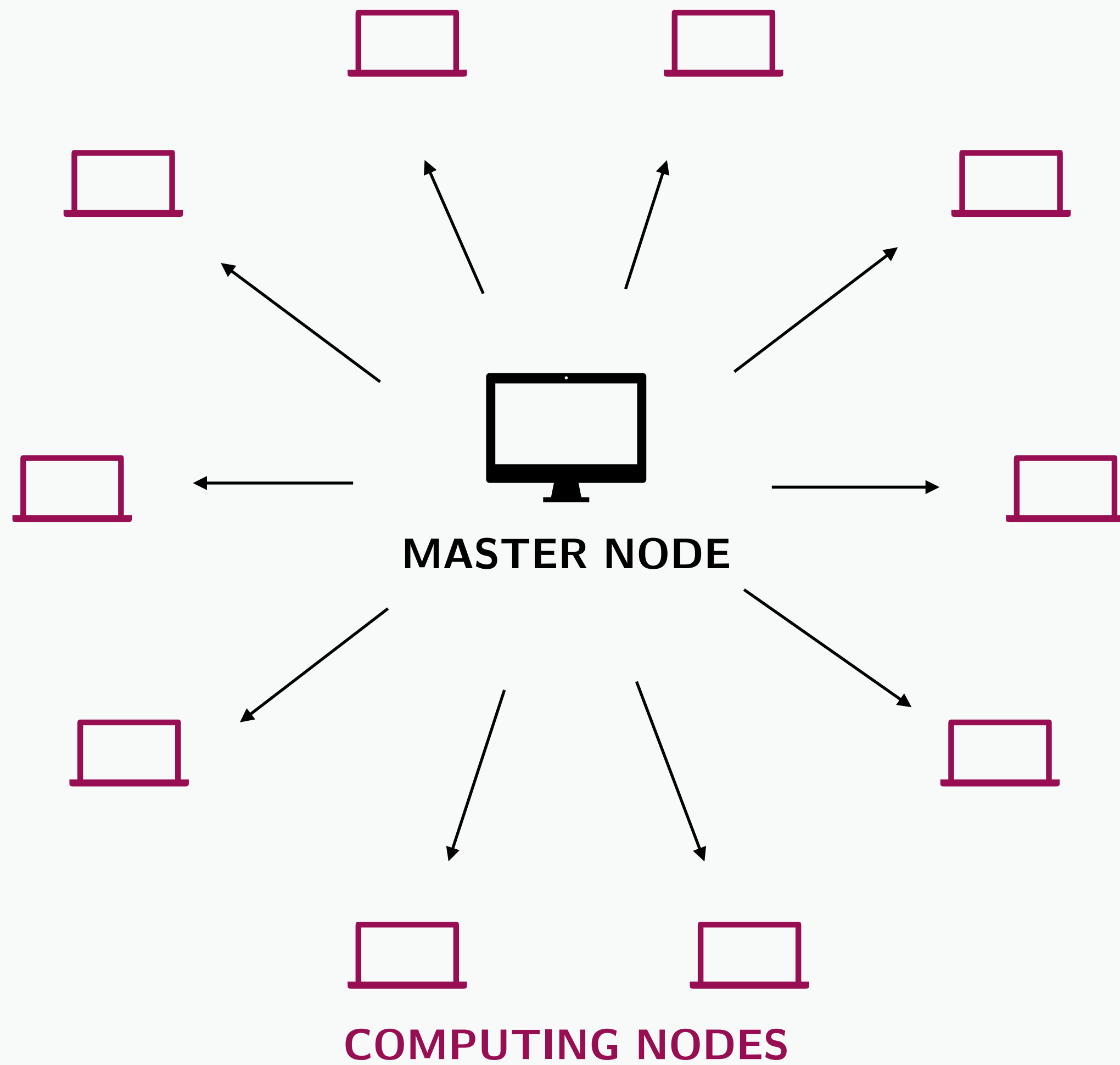


GT de l'équipe GRACE

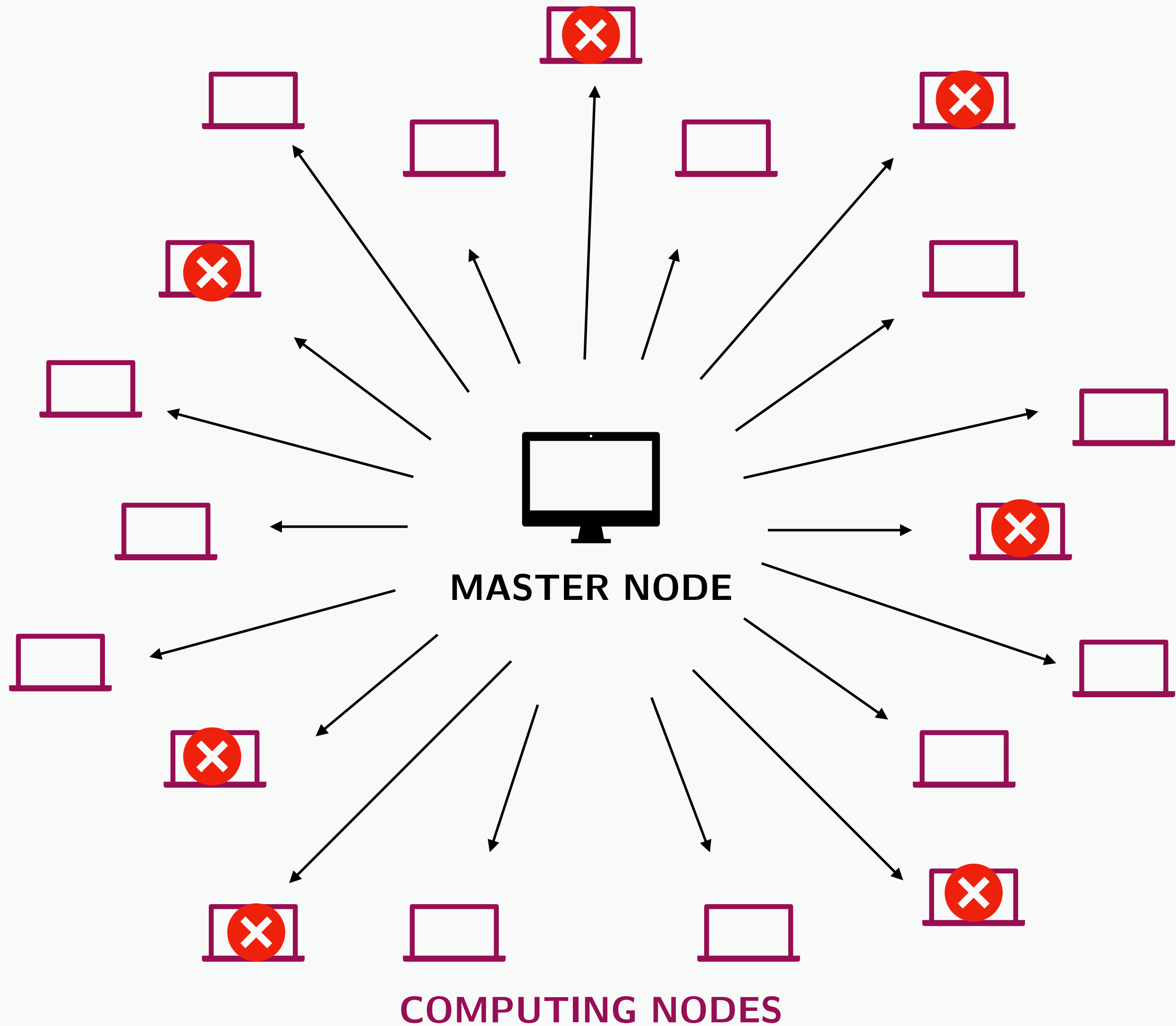
December 1, 2020

High Performance Computing Technologies

Goal: provide high performances and complete heavy tasks



Fault tolerant algorithms



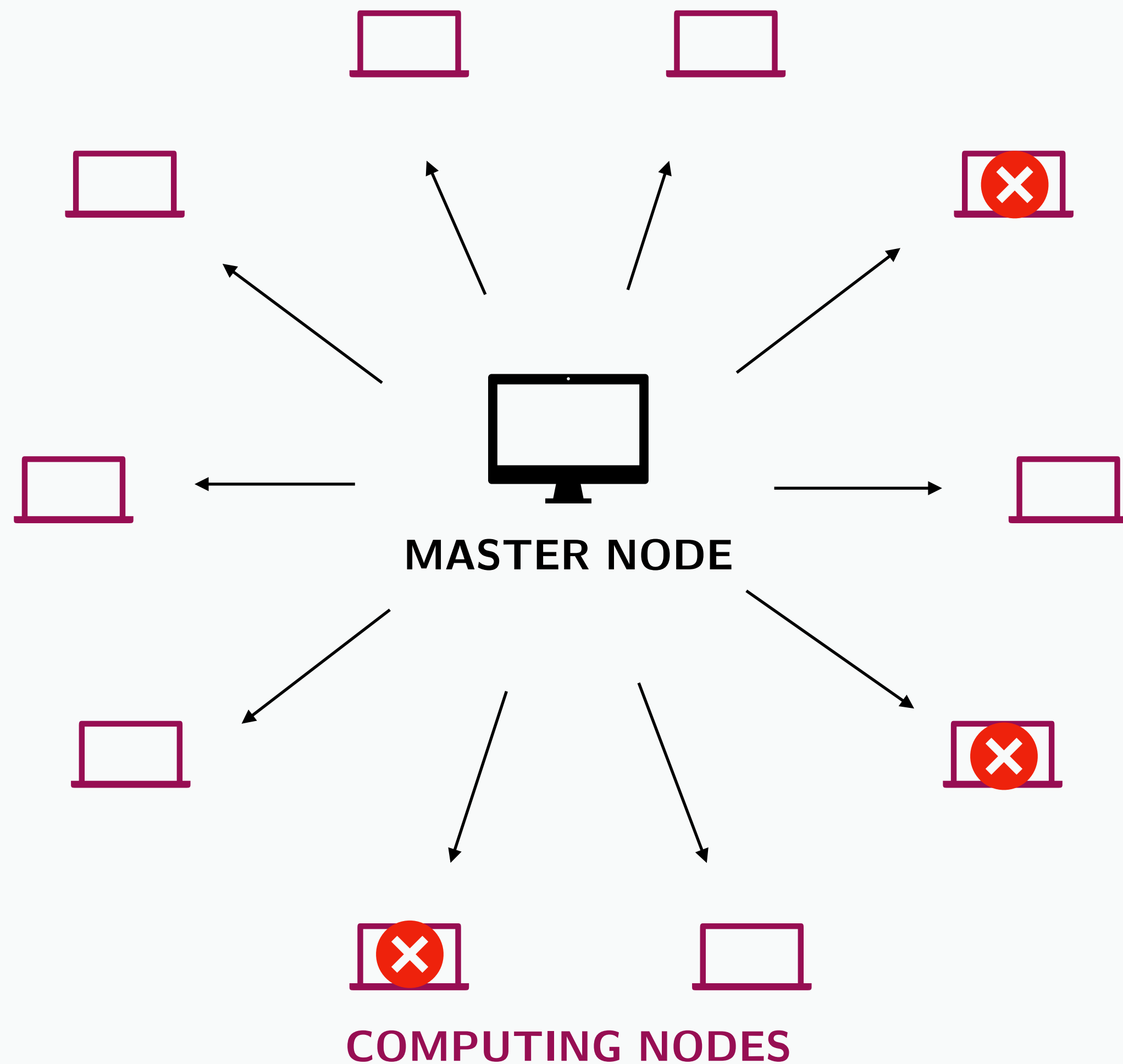
The **more the number of system components** grows
the **more the failures** of computing nodes
becomes **relevant**
(3,5 faults per day)

[DI, GUO, PERSHEY, SNIR, CAPPELLO, 2019], [LIU, CHEN, 2018]

↓
construct **fault tolerant algorithms**

↓
detect/correct faults

Algorithm-based fault tolerant techniques (ABFT)



Algorithm-based fault tolerant techniques (ABFT)

[HUANG, ABRAHAM, 1984]

exploits the **algorithm's characteristics**
to design a **fault tolerant algorithm**

Goal: detect/correct computational errors (faults)

Algorithm-based fault tolerant techniques (ABFT)

[HUANG, ABRAHAM, 1984]

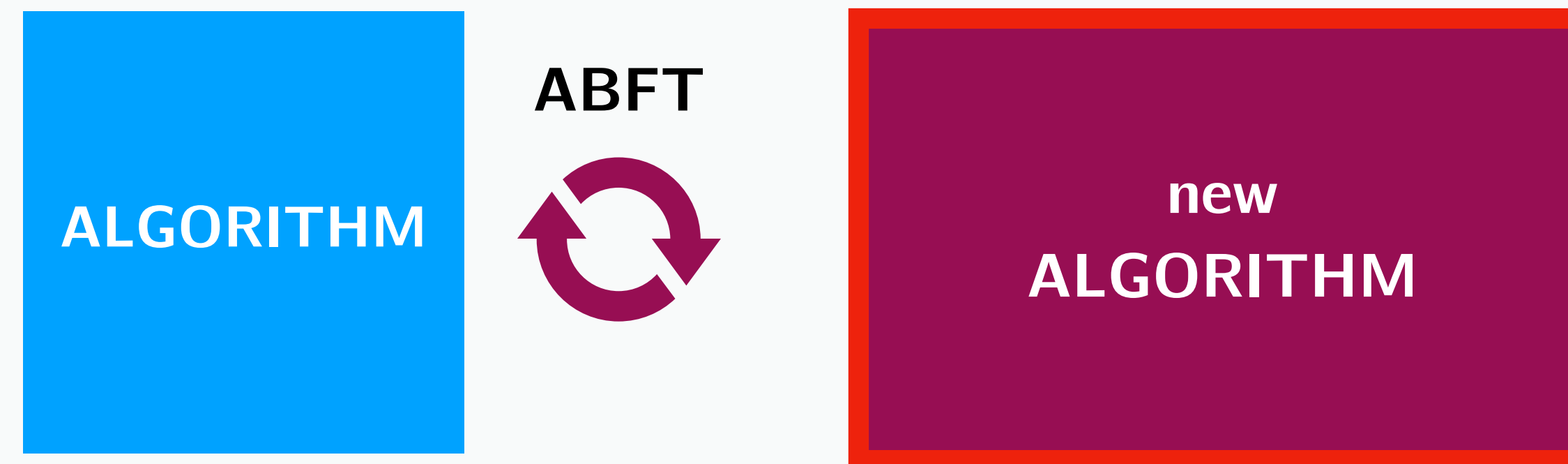
Goal: detect/correct computational errors (faults)



Algorithm-based fault tolerant techniques (ABFT)

[HUANG, ABRAHAM, 1984]

Goal: detect/correct computational errors (faults)



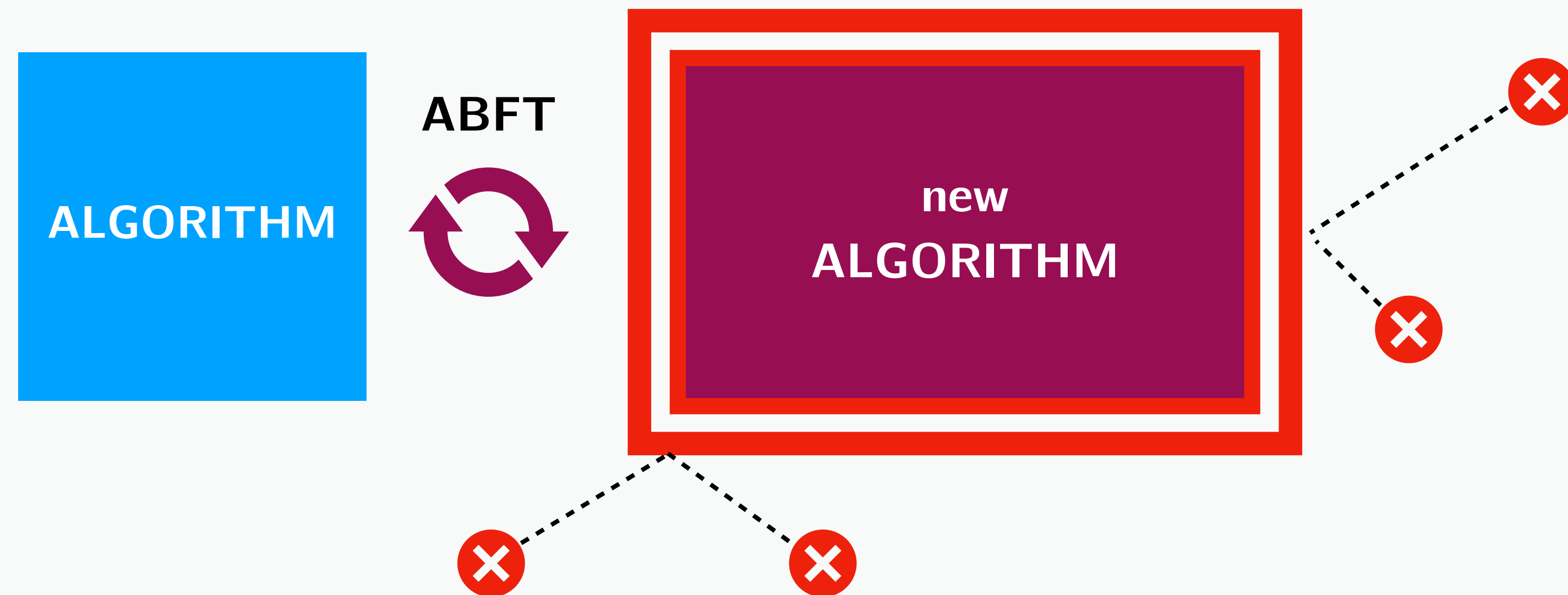
- **adding redundancy** (encoding)

High Performance Computing Technologies

Algorithm-based fault tolerant techniques (ABFT)

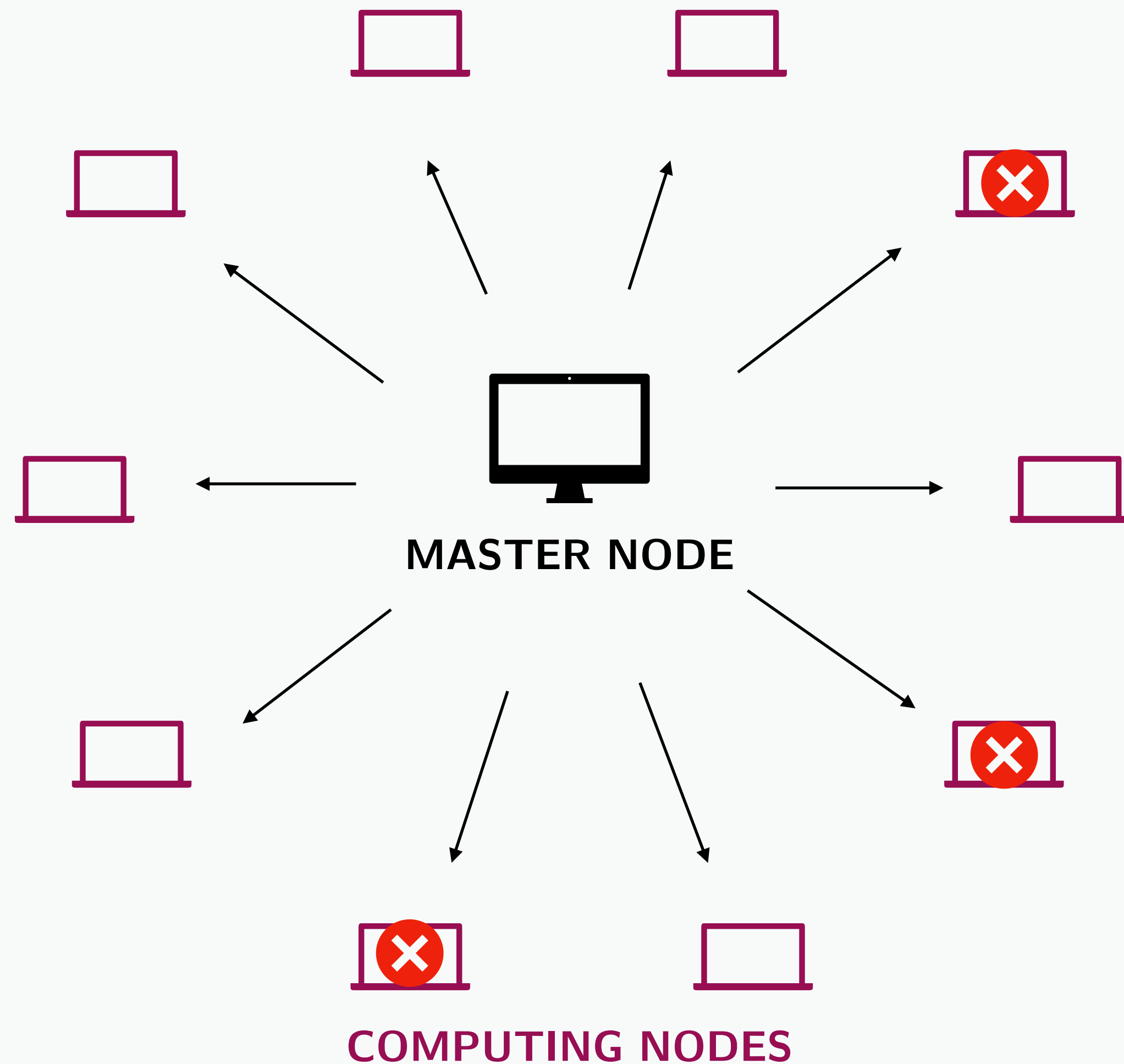
[HUANG, ABRAHAM, 1984]

Goal: detect/correct computational errors (faults)



- **adding redundancy** (encoding)
- modify the algorithm, work on encoding data ← **robust to errors (faults)**

error correcting codes



Algorithm-based fault tolerant techniques (ABFT)

[HUANG, ABRAHAM, 1984]

exploits the **algorithm's characteristics**
to design a **fault tolerant algorithm**

Goal: detect/correct computational errors (faults)

ABFT for Polynomial Linear System Solving by Evaluation-Interpolation

[BOYER, KALTOFEN, 2014]

[KALTOFEN, PERNET, STORJOHANN, WADDEL, 2017]

 [GUERRINI, LEBRETON, Z., 2019]

 [GUERRINI, LEBRETON, Z., 2020]

 : publication

 : preprint

Take a look inside



GT de l'équipe GRACE

December 1, 2020

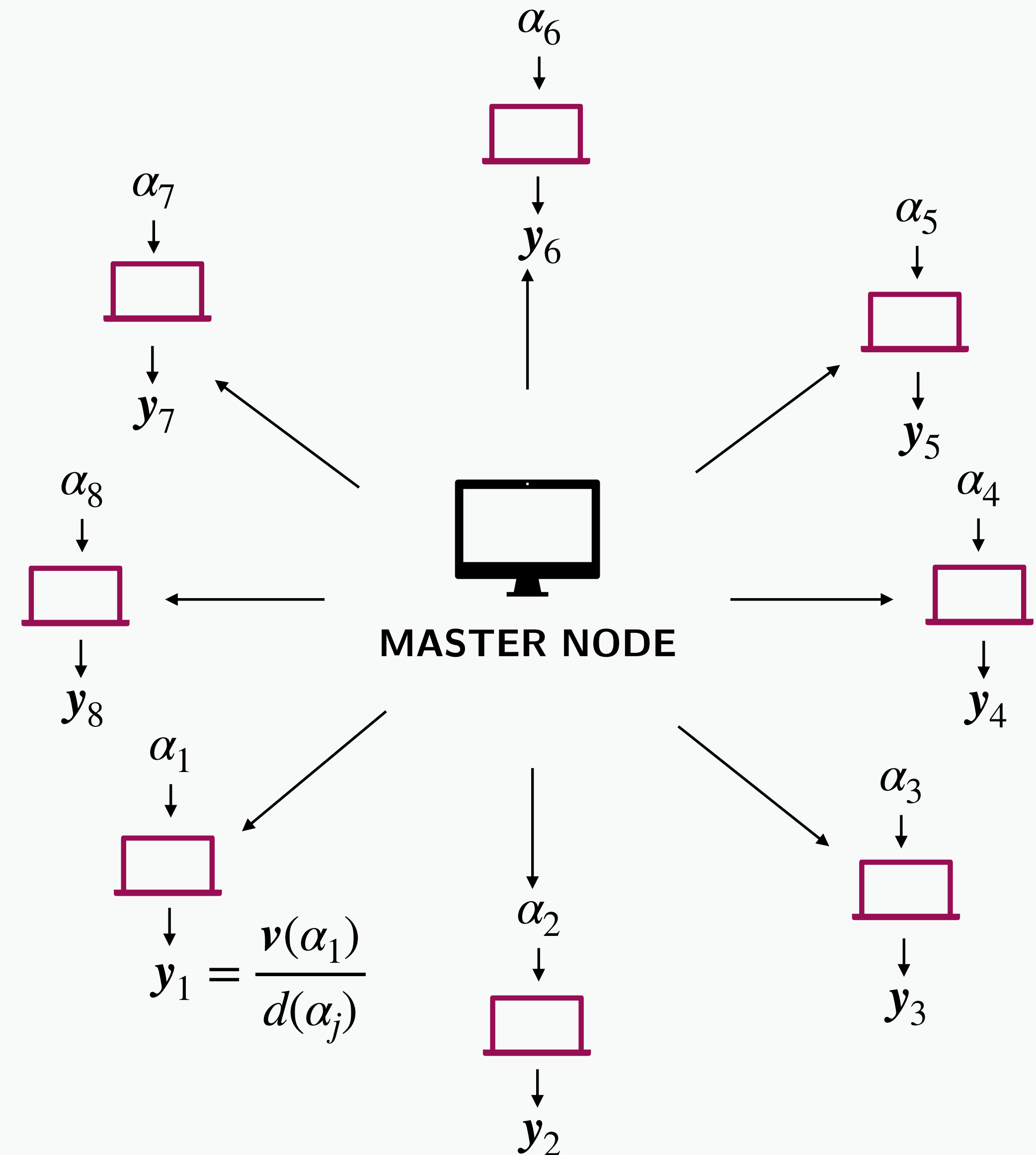
Polynomial Linear System Solving by Evaluation-Interpolation

Polynomial Linear System Solving

$$\underbrace{\begin{pmatrix} a_{1,1}(x) & a_{1,2}(x) & \dots & a_{1,n}(x) \\ a_{2,1}(x) & a_{2,2}(x) & \dots & a_{2,n}(x) \\ \vdots & \vdots & \vdots & \vdots \\ a_{n,1}(x) & a_{n,2}(x) & \dots & a_{n,n}(x) \end{pmatrix}}_{A(x)} \underbrace{\begin{pmatrix} y_1(x) \\ y_2(x) \\ \vdots \\ y_n(x) \end{pmatrix}}_{\mathbf{y}(x)} = \underbrace{\begin{pmatrix} b_1(x) \\ b_2(x) \\ \vdots \\ b_n(x) \end{pmatrix}}_{\mathbf{b}(x)}$$

Evaluation-Interpolation Algorithm

1. **Evaluate** $A(x)$, $\mathbf{b}(x)$ in $\{\alpha_1, \dots, \alpha_L\}$, ($A(\alpha_j)$ full rank)
2. **Compute** $y_j = A(\alpha_j)^{-1} \mathbf{b}(\alpha_j) = \frac{v(\alpha_j)}{d(\alpha_j)}$,
3. Interpolate $\mathbf{y}(x)$ from $\mathbf{y}_1, \dots, \mathbf{y}_L$



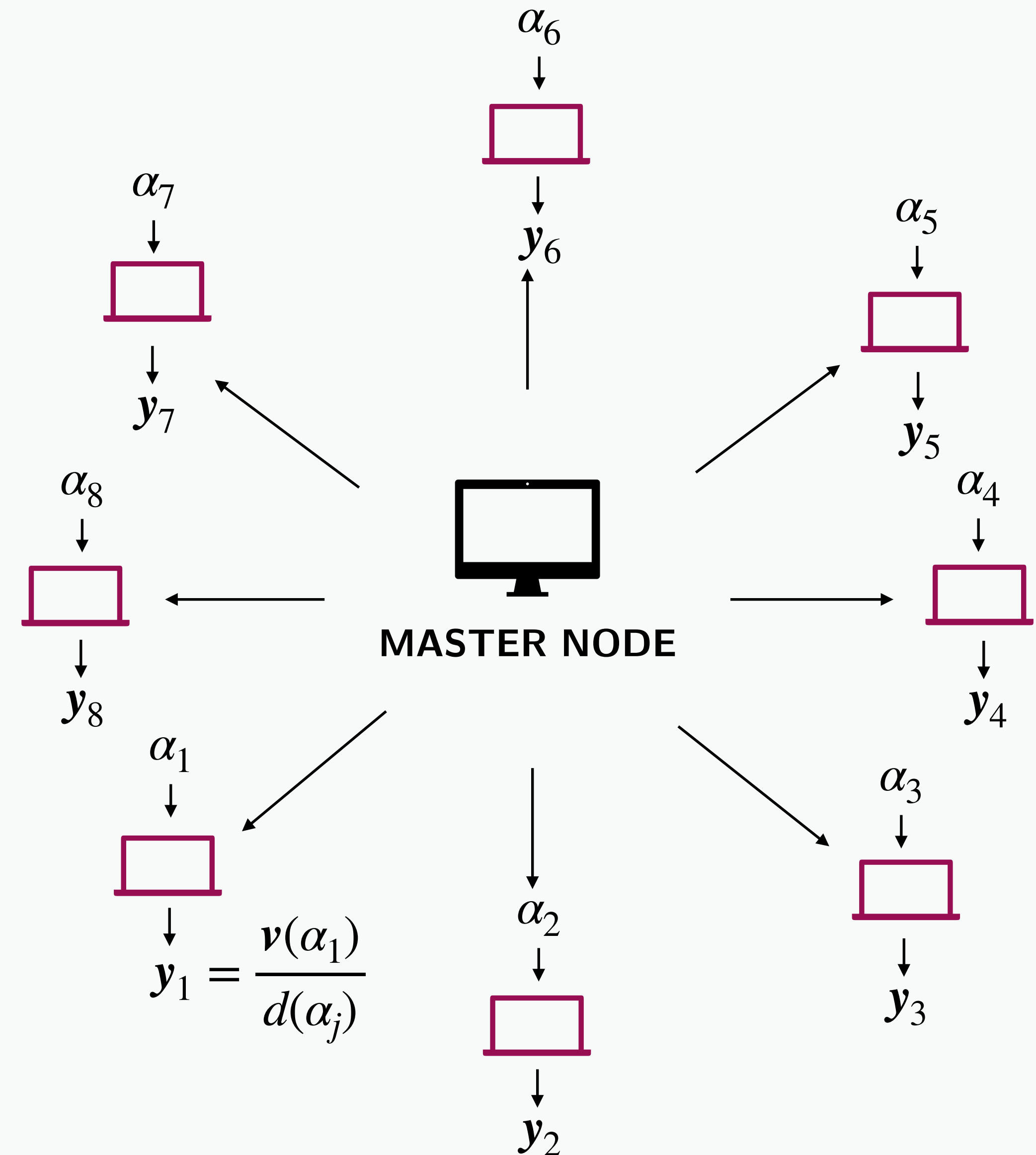
Polynomial Linear System Solving by Evaluation-Interpolation

Polynomial Linear System Solving

$$\underbrace{\begin{pmatrix} a_{1,1}(x) & a_{1,2}(x) & \dots & a_{1,n}(x) \\ a_{2,1}(x) & a_{2,2}(x) & \dots & a_{2,n}(x) \\ \vdots & \vdots & \vdots & \vdots \\ a_{n,1}(x) & a_{n,2}(x) & \dots & a_{n,n}(x) \end{pmatrix}}_{A(x)} \underbrace{\begin{pmatrix} \frac{v_1(x)}{d(x)} \\ \vdots \\ \frac{v_n(x)}{d(x)} \end{pmatrix}}_{\mathbf{y}(x)} = \underbrace{\begin{pmatrix} b_1(x) \\ b_2(x) \\ \vdots \\ b_n(x) \end{pmatrix}}_{\mathbf{b}(x)}$$

Evaluation-Interpolation Algorithm

1. **Evaluate** $A(x)$, $\mathbf{b}(x)$ in $\{\alpha_1, \dots, \alpha_L\}$, ($A(\alpha_j)$ full rank)
2. **Compute** $y_j = A(\alpha_j)^{-1} \mathbf{b}(\alpha_j) = \frac{\mathbf{v}(\alpha_j)}{d(\alpha_j)}$,
3. Interpolate $\mathbf{y}(x)$ from $\mathbf{y}_1, \dots, \mathbf{y}_L$



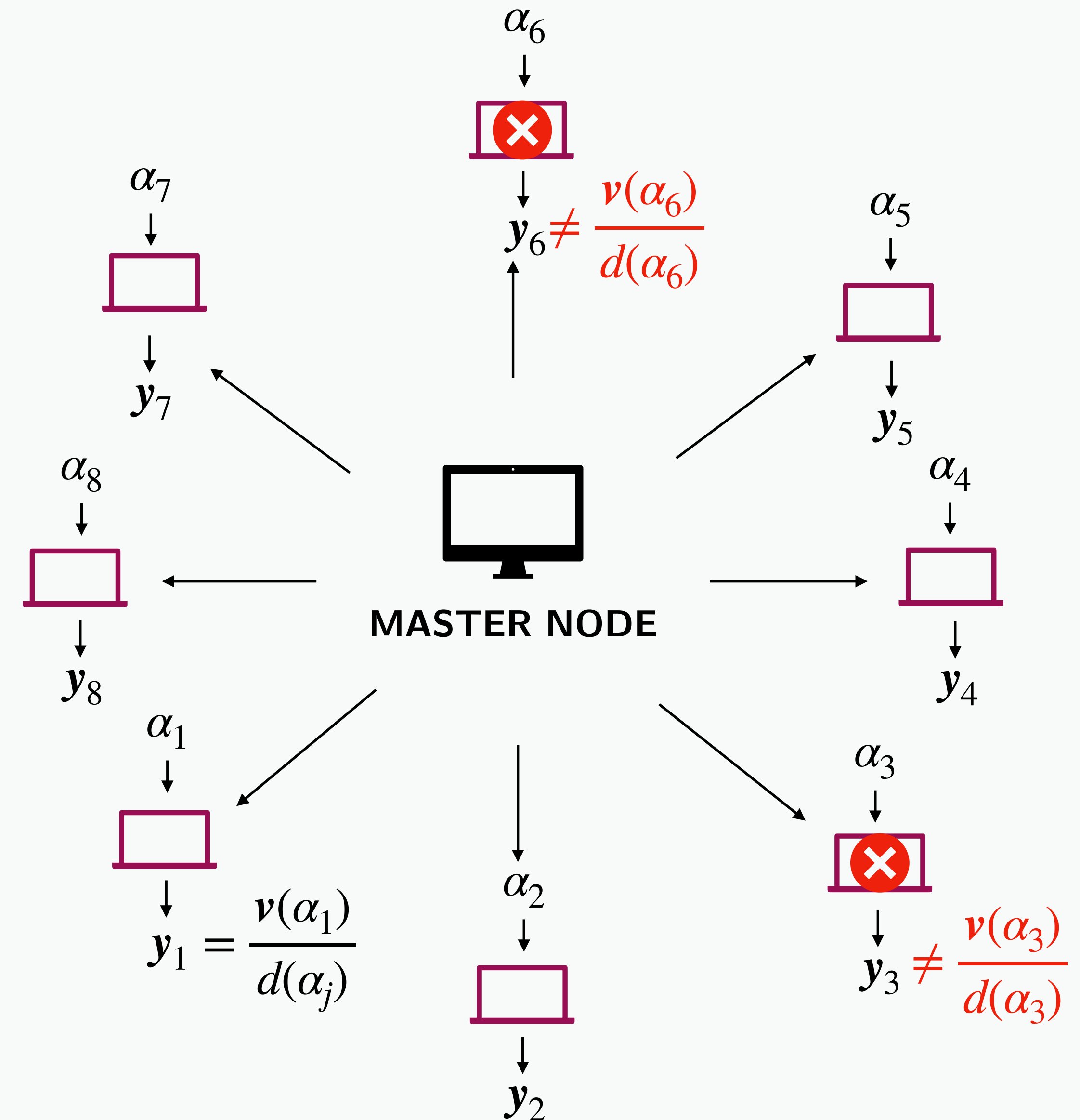
Polynomial Linear System Solving by Evaluation-Interpolation

Polynomial Linear System Solving

$$\underbrace{\begin{pmatrix} a_{1,1}(x) & a_{1,2}(x) & \dots & a_{1,n}(x) \\ a_{2,1}(x) & a_{2,2}(x) & \dots & a_{2,n}(x) \\ \vdots & \vdots & \vdots & \vdots \\ a_{n,1}(x) & a_{n,2}(x) & \dots & a_{n,n}(x) \end{pmatrix}}_{A(x)} \underbrace{\begin{pmatrix} \frac{v_1(x)}{d(x)} \\ \vdots \\ \frac{v_n(x)}{d(x)} \end{pmatrix}}_{\mathbf{y}(x)} = \underbrace{\begin{pmatrix} b_1(x) \\ b_2(x) \\ \vdots \\ b_n(x) \end{pmatrix}}_{\mathbf{b}(x)}$$

Evaluation-Interpolation Algorithm

1. **Evaluate** $A(x)$, $\mathbf{b}(x)$ in $\{\alpha_1, \dots, \alpha_L\}$, ($A(\alpha_j)$ full rank)
2. **Compute** $y_j = A(\alpha_j)^{-1} \mathbf{b}(\alpha_j) = \frac{v(\alpha_j)}{d(\alpha_j)}$,
3. Interpolate $\mathbf{y}(x)$ from $\mathbf{y}_1, \dots, \mathbf{y}_L$



ABFT for PLS solving by Evaluation-Interpolation

ALGORITHM

ABFT

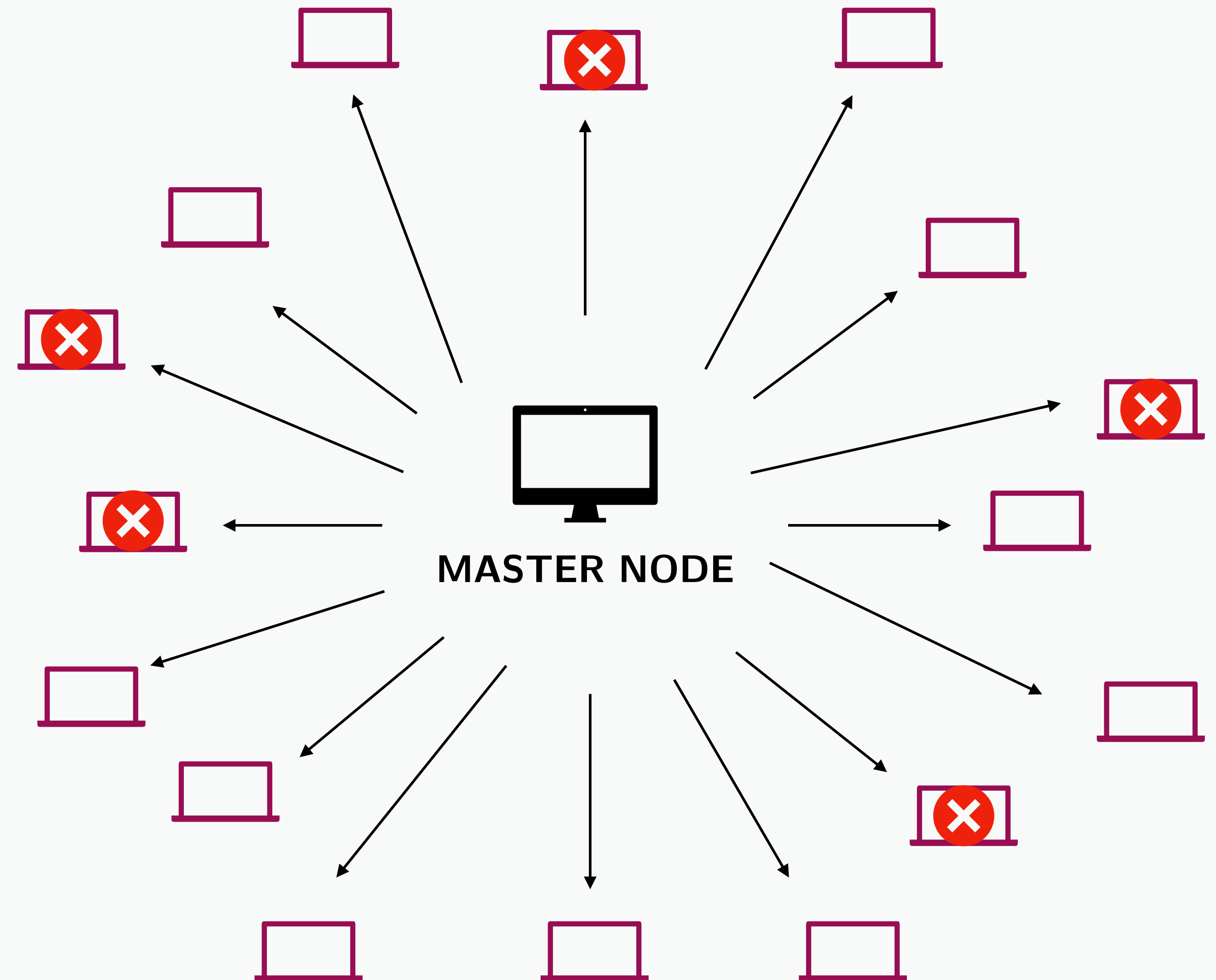


new
ALGORITHM

- **adding** redundancy (encoding), consider many nodes
- decoding, correcting errors

ABFT for Evaluation-Interpolation Algorithm

1. **Evaluate** $A(x)$, $\mathbf{b}(x)$ in many evaluation points
2. **Compute** $\mathbf{y}_j = A(\alpha_j)^{-1}\mathbf{b}(\alpha_j)$
3. Interpolate $\mathbf{y}(x)$ from $\mathbf{y}_1, \dots, \mathbf{y}_L$ where some errors occur



ABFT for PLS solving by Evaluation-Interpolation

ALGORITHM



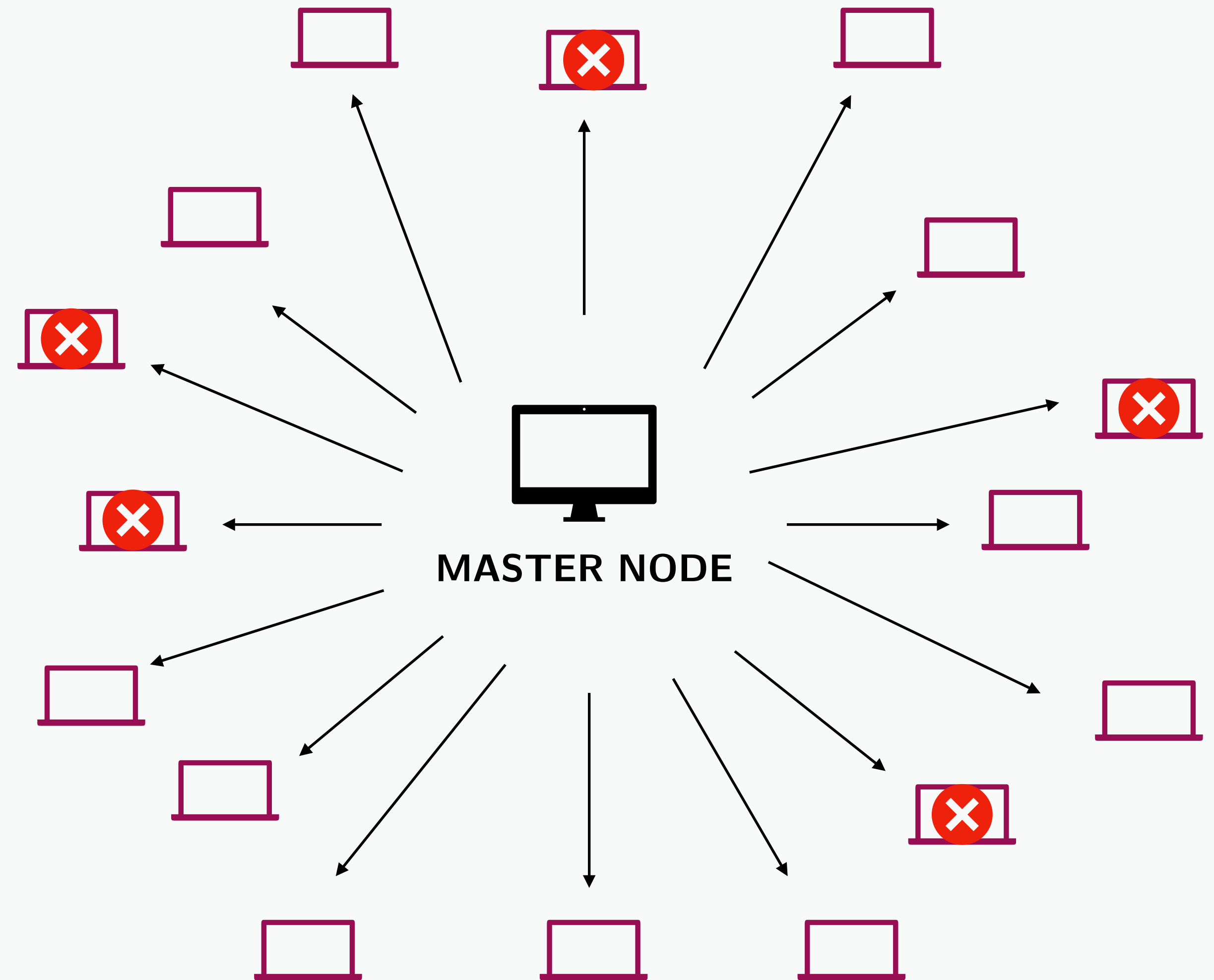
new
ALGORITHM

- **adding** redundancy (encoding), consider many nodes
- **decoding**, correcting errors

ABFT for Evaluation-Interpolation Algorithm

1. Evaluate $A(x)$, $b(x)$ in many evaluation points
2. Compute $y_j = A(\alpha_j)^{-1}b(\alpha_j)$
3. Interpolate $y(x)$ from y_1, \dots, y_L where some errors occur

Simultaneous Cauchy Interpolation with Errors



Simultaneous Cauchy Interpolation with Errors

Simultaneous Cauchy Interpolation with Errors

Given $\mathbf{y}_1, \dots, \mathbf{y}_L$

- $\mathbf{y}_j = \frac{\mathbf{v}(\alpha_j)}{d(\alpha_j)}$ correct evaluations
- $\mathbf{y}_j \neq \frac{\mathbf{v}(\alpha_j)}{d(\alpha_j)}$ erroneous evaluations

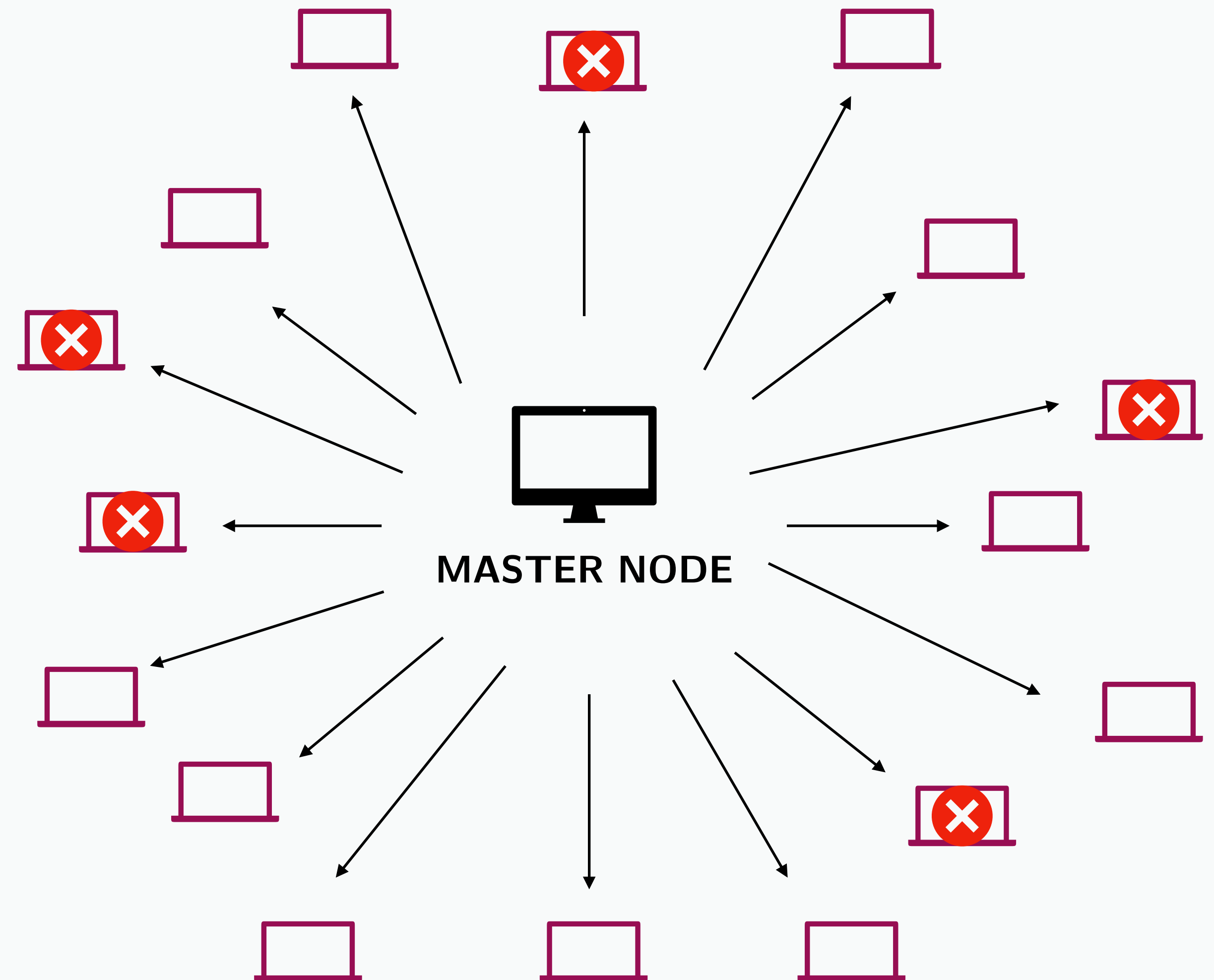
the degree bounds $N > \deg(\mathbf{v})$, $D > \deg(d)$
and an upper bound τ on the number of errors.

GOAL: reconstruct $(\mathbf{v}(x), d(x)) \rightarrow \mathbf{y}(x)$



Simultaneous Cauchy Interpolation

Simultaneous Rational Function Reconstruction






Two Main Questions

1. How many evaluations (nodes) do we need to uniquely recover $(v(x), d(x))$?




fewer evaluations \longrightarrow fewer computations

2. Can we reduce this number?

Number of Evaluations - Outline of this work

			uniqueness	uniqueness <i>almost always</i>
no-errors	$(\mathbf{v}(x), d(x))$		Cauchy Interpolation	 [GUERRINI, LEBRETON, Z., 2020]
	$A(x)\frac{\mathbf{v}(x)}{d(x)} = \mathbf{b}(x)$		[CABAY, 1971]	?
with errors	$(\mathbf{v}(x), d(x))$	d constant ($D = 0$)	Unique Decoding Capability Theorem (IRS codes)	IRS codes [BLEICHENBACHER, KIAYIAS, YUNG, 2003]
			[BOYER, KALTOFEN, 2014]	 [GUERRINI, LEBRETON, Z., 2019]
	$A(x)\frac{\mathbf{v}(x)}{d(x)} = \mathbf{b}(x)$	$D > 0$	[KALTOFEN, PERNET, STORJOHANN, WADDEL, 2017]	 [GUERRINI, LEBRETON, Z., 2020]

Number of Evaluations - Outline of this work

			uniqueness	uniqueness <i>almost always</i>
no-errors	$(\mathbf{v}(x), d(x))$		Cauchy Interpolation	 [GUERRINI, LEBRETON, Z., 2020]
	$A(x) \frac{\mathbf{v}(x)}{d(x)} = \mathbf{b}(x)$		[CABAY, 1971]	?
with errors	$(\mathbf{v}(x), d(x))$	d constant ($D = 0$)	Unique Decoding Capability Theorem (IRS codes)	IRS codes [BLEICHENBACHER, KIAYIAS, YUNG, 2003]
			[BOYER, KALTOFEN, 2014]	 [GUERRINI, LEBRETON, Z., 2019]
	$A(x) \frac{\mathbf{v}(x)}{d(x)} = \mathbf{b}(x)$	$D > 0$	[KALTOFEN, PERNET, STORJOHANN, WADDEL, 2017]	 [GUERRINI, LEBRETON, Z., 2020]

Simultaneous Cauchy Interpolation

Given the vectors $\mathbf{y}_1, \dots, \mathbf{y}_L$
and the degree bounds N, D

GOAL: **find** $(\underbrace{v_1(x), \dots, v_n(x)}_{v(x)}, d(x))$ s.t.

- $v_i(\alpha_j) = y_{i,j}d(\alpha_j)$
- $\deg(v_i) < N$
- $\deg(d) < D$



Simultaneous Rational Function Reconstruction

Simultaneous Cauchy Interpolation

Given the vectors $\mathbf{y}_1, \dots, \mathbf{y}_L$
and the degree bounds N, D

GOAL: **find** $(\underbrace{v_1(x), \dots, v_n(x)}_{v(x)}, d(x))$ s.t.

- $v_i(\alpha_j) = y_{i,j}d(\alpha_j)$
- $\deg(v_i) < N$
- $\deg(d) < D$



Simultaneous Rational Function Reconstruction



Vector Generalization
same denominator

Cauchy Interpolation

Given the evaluations y_1, \dots, y_L
and the degree bounds N, D

GOAL: **find** $(v(x), d(x))$ s.t.

- $\frac{v(\alpha_j)}{d(\alpha_j)} = y_j, d(\alpha_j) \neq 0$
- $\deg(v) < N$
- $\deg(d) < D$

No-error case

Simultaneous Cauchy Interpolation

Given the vectors $\mathbf{y}_1, \dots, \mathbf{y}_L$
and the degree bounds N, D

GOAL: **find** $(\underbrace{v_1(x), \dots, v_n(x)}_{v(x)}, d(x))$ s.t.

- $v_i(\alpha_j) = y_{i,j}d(\alpha_j)$
- $\deg(v_i) < N$
- $\deg(d) < D$



Simultaneous Rational Function Reconstruction



Vector Generalization
same denominator

Cauchy Interpolation

Given the evaluations y_1, \dots, y_L
and the degree bounds N, D

GOAL: **find** $(v(x), d(x))$ s.t.

- $\frac{v(\alpha_j)}{d(\alpha_j)} = y_j, d(\alpha_j) \neq 0$
- $\deg(v) < N$
- $\deg(d) < D$



Rational Function Reconstruction

If the number of evaluations $L \geq N + D - 1$



Unique solution

$(v_1, d_1), (v_2, d_2)$ solutions $\longrightarrow \frac{v_1}{d_1} = \frac{v_2}{d_2}$

No-error case

Simultaneous Cauchy Interpolation

Given the vectors $\mathbf{y}_1, \dots, \mathbf{y}_L$
and the degree bounds N, D

GOAL: **find** $(\underbrace{v_1(x), \dots, v_n(x)}_{v(x)}, d(x))$ s.t.

- $v_i(\alpha_j) = y_{i,j}d(\alpha_j)$
- $\deg(v_i) < N$
- $\deg(d) < D$



Simultaneous Rational Function Reconstruction



Vector Generalization
same denominator

Cauchy Interpolation

Given the evaluations y_1, \dots, y_L
and the degree bounds N, D

GOAL: **find** $(v(x), d(x))$ s.t.

- $v(\alpha_j) = y_jd(\alpha_j)$
- $\deg(v) < N$
- $\deg(d) < D$



Rational Function Reconstruction

If the number of evaluations $L \geq N + D - 1$



Unique solution

$(v_1, d_1), (v_2, d_2)$ solutions $\longrightarrow \frac{v_1}{d_1} = \frac{v_2}{d_2}$

No-error case

Simultaneous Cauchy Interpolation

Given the vectors $\mathbf{y}_1, \dots, \mathbf{y}_L$
and the degree bounds N, D

GOAL: **find** $(v_1(x), \dots, v_n(x), d(x))$ s.t.

- $\underbrace{\hspace{10em}}_{v(x)}$
- $v_i(\alpha_j) = y_{i,j}d(\alpha_j)$
 - $\deg(v_i) < N$
 - $\deg(d) < D$



Vector Generalization
same denominator

Cauchy Interpolation

Given the evaluations y_1, \dots, y_L
and the degree bounds N, D

GOAL: **find** $(v(x), d(x))$ s.t.

- $v(\alpha_j) = y_jd(\alpha_j)$
- $\deg(v) < N$
- $\deg(d) < D$

1. Apply the **Cauchy Interpolation** component-wise ($L \geq N + D - 1 \longrightarrow$ **uniqueness**)

No-error case

Simultaneous Cauchy Interpolation

Given the vectors $\mathbf{y}_1, \dots, \mathbf{y}_L$
and the degree bounds N, D

GOAL: **find** $(v_1(x), \dots, v_n(x), d(x))$ s.t.

- $\underbrace{\hspace{10em}}_{v(x)}$
- $v_i(\alpha_j) = y_{i,j}d(\alpha_j)$
 - $\deg(v_i) < N$
 - $\deg(d) < D$

Cauchy Interpolation

Given the evaluations y_1, \dots, y_L
and the degree bounds N, D

GOAL: **find** $(v(x), d(x))$ s.t.

- $v(\alpha_j) = y_jd(\alpha_j)$
- $\deg(v) < N$
- $\deg(d) < D$



Vector Generalization
same denominator

1. Apply the **Cauchy Interpolation** component-wise ($L \geq N + D - 1 \longrightarrow$ **uniqueness**)
2. Use the **common denominator feature** to reduce the number of equations of the related **homogeneous linear system**

$$\begin{array}{ccc} \swarrow & \# \text{equations} = \# \text{unknowns} - 1 \longrightarrow & L = N + (D - 1)/n \\ & nL & nN + D & \text{uniqueness?} \end{array}$$

No-error case

Simultaneous Cauchy Interpolation

Given the vectors $\mathbf{y}_1, \dots, \mathbf{y}_L$
and the degree bounds N, D

GOAL: **find** $(\underbrace{v_1(x), \dots, v_n(x)}_{v(x)}, d(x))$ s.t.

- $v_i(\alpha_j) = y_{i,j}d(\alpha_j)$
- $\deg(v_i) < N$
- $\deg(d) < D$



Vector Generalization
same denominator

Cauchy Interpolation

Given the evaluations y_1, \dots, y_L
and the degree bounds N, D

GOAL: **find** $(v(x), d(x))$ s.t.

- $v(\alpha_j) = y_jd(\alpha_j)$
- $\deg(v) < N$
- $\deg(d) < D$

If we want to recover a solution of a PLS:

- with $L \geq \max\{\deg(A) + N, \deg(\mathbf{b}) + D\} \longrightarrow$ **uniqueness** [CABAY, 1971]

No-error case

Simultaneous Cauchy Interpolation

Given the vectors $\mathbf{y}_1, \dots, \mathbf{y}_L$
and the degree bounds N, D

GOAL: **find** $(\underbrace{v_1(x), \dots, v_n(x)}_{v(x)}, d(x))$ s.t.

- $v_i(\alpha_j) = y_{i,j}d(\alpha_j)$
- $\deg(v_i) < N$
- $\deg(d) < D$



Vector Generalization
same denominator

Cauchy Interpolation

Given the evaluations y_1, \dots, y_L
and the degree bounds N, D

GOAL: **find** $(v(x), d(x))$ s.t.

- $v(\alpha_j) = y_jd(\alpha_j)$
- $\deg(v) < N$
- $\deg(d) < D$

If we want to recover a solution of a PLS:

- with $L \geq \max\{\deg(A) + N, \deg(\mathbf{b}) + D\} \longrightarrow$ **uniqueness** [CABAY, 1971]
- for **specific degree constraints** N, D $\max\{\deg(A) + N, \deg(\mathbf{b}) + D\} = N + (D - 1)/n \longrightarrow$ **uniqueness** [OLESH, STORJOHANN, 2007]

No-error case

Simultaneous Cauchy Interpolation

Given the vectors $\mathbf{y}_1, \dots, \mathbf{y}_L$
and the degree bounds N, D

GOAL: **find** $(\underbrace{v_1(x), \dots, v_n(x)}_{v(x)}, d(x))$ s.t.

- $v_i(\alpha_j) = y_{i,j}d(\alpha_j)$
- $\deg(v_i) < N$
- $\deg(d) < D$



Vector Generalization
same denominator

Cauchy Interpolation

Given the evaluations y_1, \dots, y_L
and the degree bounds N, D

GOAL: **find** $(v(x), d(x))$ s.t.

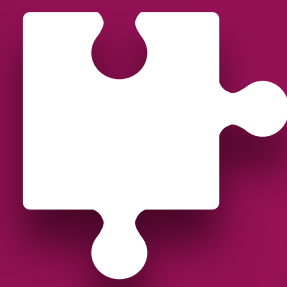
- $v(\alpha_j) = y_jd(\alpha_j)$
- $\deg(v) < N$
- $\deg(d) < D$

 **Theorem** [GUERRINI, LEBRETON, Z., 2020]

If $L = N + (D - 1)/n$, for almost all instances \implies uniqueness.

If $\mathbb{K} = \mathbb{F}_q$, the proportion of instances leading to non-uniqueness is $\leq (D - 1)/q$

Just a hint of the technique...



GT de l'équipe GRACE

December 1, 2020

Simultaneous Cauchy Interpolation

Given the vectors $\mathbf{y}_1, \dots, \mathbf{y}_L$
and the degree bounds N, D

GOAL: **find** $(v_1(x), \dots, v_n(x), d(x))$ s.t.

- $\underbrace{\hspace{10em}}_{v(x)}$
- $v_i(\alpha_j) = y_{i,j}d(\alpha_j)$
 - $\deg(v_i) < N$
 - $\deg(d) < D$

 **Theorem** [GUERRINI, LEBRETON, Z., 2020]

If $L = \deg(\mathbf{a}) = N + (D - 1)/n$, for almost all instances \implies uniqueness.

If $\mathbb{K} = \mathbb{F}_q$, the proportion of instances leading to non-uniqueness is $\leq (D - 1)/q$

Simultaneous Cauchy Interpolation

Given the vectors $\mathbf{y}_1, \dots, \mathbf{y}_L$
and the degree bounds N, D

GOAL: **find** $(v_1(x), \dots, v_n(x), d(x))$ s.t.

- $v_i(x) = u_i(x)d(x) \bmod \prod (x - \alpha_j)$
- $\deg(v_i) < N$
- $\deg(d) < D$ $\mathbf{u}(x)$ vector of Lagrange interpolators

→
Specific case of

Simultaneous Rational Function Reconstruction

Given two vector of polynomials $\mathbf{u}(x), \mathbf{a}(x)$
and the degree bounds N, D

GOAL: **find** $(v_1(x), \dots, v_n(x), d(x))$ s.t.

- $v_i(x) = u_i(x)d(x) \bmod a_i(x)$
- $\deg(v_i) < N$
- $\deg(d) < D$

Theorem [GUERRINI, LEBRETON, Z., 2020]

If $L = \deg(\mathbf{a}) = N + (D - 1)/n$, for almost all instances \implies uniqueness.

If $\mathbb{K} = \mathbb{F}_q$, the proportion of instances leading to non-uniqueness is $\leq (D - 1)/q$

No-error case

Simultaneous Rational Function Reconstruction

Given two vector of polynomials $\mathbf{u}(x), \mathbf{a}(x)$
and the degree bounds N, D

GOAL: find $(v_1(x), \dots, v_n(x), d(x))$ s.t.

- $v_i(x) = u_i(x)d(x) \bmod a_i(x)$
- $\deg(v_i) < N$
- $\deg(d) < D$

$$(\mathbf{v}, d) \underbrace{\begin{pmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \\ -u_1 & -u_2 & \dots & -u_n \end{pmatrix}}_{R_u} = 0 \bmod \underbrace{\langle (0, \dots, a_i, \dots, 0) \rangle}_{\mathcal{M}} \iff (\mathbf{v}, d) \in \mathcal{A}_{R_u}$$

$$\mathcal{A}_{R_u} = \{ \mathbf{p} = (p_1(x), \dots, p_n(x)) \mid \mathbf{p}R_u = \mathbf{0} \bmod \mathcal{M} \}$$

Relation module

 **Theorem** [GUERRINI, LEBRETON, Z., 2020]

If $L = \deg(\mathbf{a}) = N + (D - 1)/n$, for almost all instances \implies uniqueness.

If $\mathbb{K} = \mathbb{F}_q$, the proportion of instances leading to non-uniqueness is $\leq (D - 1)/q$

Simultaneous Rational Function Reconstruction

Given two vector of polynomials $\mathbf{u}(x), \mathbf{a}(x)$
and the degree bounds N, D

GOAL: find $(v_1(x), \dots, v_n(x), d(x))$ s.t.

- $v_i(x) = u_i(x)d(x) \bmod a_i(x)$
- $\deg(v_i) < N$
- $\deg(d) < D$

$$(\mathbf{v}, d) \begin{pmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \\ -u_1 & -u_2 & \dots & -u_n \end{pmatrix} = 0 \bmod \langle (0, \dots, a_i, \dots, 0) \rangle \iff (\mathbf{v}, d) \in \mathcal{A}_{R_u}$$

take the max

$$(\deg(v_1) - N, \dots, \deg(v_n) - N, \deg(d) - D) \xrightarrow{\text{take the max}} rdeg_{(-N, \dots, -N, -D)}(\mathbf{v}, d) < 0$$

< 0 < 0 < 0

Theorem [GUERRINI, LEBRETON, Z., 2020]

If $L = \deg(\mathbf{a}) = N + (D - 1)/n$, for almost all instances \implies uniqueness.

If $\mathbb{K} = \mathbb{F}_q$, the proportion of instances leading to non-uniqueness is $\leq (D - 1)/q$

Simultaneous Rational Function Reconstruction

Given two vector of polynomials $\mathbf{u}(x), \mathbf{a}(x)$
and the degree bounds N, D

GOAL: find $(\underbrace{v_1(x), \dots, v_n(x)}_{\mathbf{v}(x)}, d(x))$ s.t.

- $(\mathbf{v}, d) \in \mathcal{A}_{R_u}$
- $rdeg_{(-N, \dots, -N, -D)}(\mathbf{v}, d) < 0$

How to prove uniqueness?

- **Minimal basis** \mathcal{B} of \mathcal{A}_{R_u} , for which the s -row degrees are uniquely defined
(Ordered Weak Popov)
- **Solution space generated** by elements of \mathcal{B} the with **negative** s -row degrees

↓
 $(-N, \dots, -N, -D)$ -row degrees of \mathcal{B} of the form $(0, 0, \dots, -1)$




↓
Solution space uniquely generated \Rightarrow uniqueness

Theorem [GUERRINI, LEBRETON, Z., 2020]




If $L = \deg(\mathbf{a}) = N + (D - 1)/n$, for almost all instances, $rdeg_{(-N, \dots, -N, -D)}(\mathcal{B}) = (0, \dots, 0, -1)$

If $\mathbb{K} = \mathbb{F}_q$, the **proportion of instances** leading to **non-uniqueness** is $\leq (D - 1)/q$

Number of Evaluations - Outline of this work

			uniqueness	uniqueness <i>almost always</i>
no-errors	$(\mathbf{v}(x), d(x))$		$L = N + D - 1$ Cauchy Interpolation	 $L = N + \left\lceil \frac{(D-1)}{n} \right\rceil$
	$A(x) \frac{\mathbf{v}(x)}{d(x)} = \mathbf{b}(x)$		$L = \max\{\deg(A) + N, \deg(\mathbf{b}) + D\}$ [CABAY, 1971]	
with errors	$(\mathbf{v}(x), d(x))$	d constant ($D = 0$)	Unique Decoding Capability Theorem (IRS codes)	IRS codes [BLEICHENBACHER, KIAYIAS, YUNG, 2003]
			[BOYER, KALTOFEN, 2014]	 [GUERRINI, LEBRETON, Z., 2019]
	$A(x) \frac{\mathbf{v}(x)}{d(x)} = \mathbf{b}(x)$	$D > 0$	[KALTOFEN, PERNET, STORJOHANN, WADDEL, 2017]	 [GUERRINI, LEBRETON, Z., 2020]

Number of Evaluations - Outline of this work

			uniqueness	uniqueness <i>almost always</i>
no-errors	$(\mathbf{v}(x), d(x))$		$L = N + D - 1$ Cauchy Interpolation	 $L = N + \left\lceil \frac{(D-1)}{n} \right\rceil$
	$A(x) \frac{\mathbf{v}(x)}{d(x)} = \mathbf{b}(x)$		$L = \max\{\deg(A) + N, \deg(\mathbf{b}) + D\}$ [CABAY, 1971]	
with errors	$(\mathbf{v}(x), d(x))$	d constant ($D = 0$)	Unique Decoding Capability Theorem (IRS codes)	IRS codes [BLEICHENBACHER, KIAYIAS, YUNG, 2003]
			[BOYER, KALTOFEN, 2014]	 [GUERRINI, LEBRETON, Z., 2019]
	$A(x) \frac{\mathbf{v}(x)}{d(x)} = \mathbf{b}(x)$	$D > 0$	[KALTOFEN, PERNET, STORJOHANN, WADDEL, 2017]	 [GUERRINI, LEBRETON, Z., 2020]

Simultaneous Cauchy Interpolation with Errors

Simultaneous Cauchy Interpolation with Errors

Given $\mathbf{y}_1, \dots, \mathbf{y}_L$

- $\mathbf{y}_j = \frac{\mathbf{v}(\alpha_j)}{d(\alpha_j)}$ correct evaluations
- $\mathbf{y}_j \neq \frac{\mathbf{v}(\alpha_j)}{d(\alpha_j)}$ erroneous evaluations

the degree bounds $N > \deg(\mathbf{v})$, $D > \deg(d)$
and an upper bound τ on the number of errors.

GOAL: reconstruct $(\mathbf{v}(x), d(x)) \rightarrow \mathbf{y}(x)$

Simultaneous Interpolation with Errors (D=0)

Simultaneous Interpolation with Errors

Given y_1, \dots, y_L

- $y_j = v(\alpha_j)$ correct evaluations
- $y_j \neq v(\alpha_j)$ erroneous evaluations

the degree bounds $N > \deg(v)$

and an upper bound τ on the number of errors.

GOAL: reconstruct $v(x)$

What happens if the denominator is constant?

GOAL: reconstruct a **vector of polynomials** given its **evaluations**,
some of which are erroneous



decoding **Interleaved Reed-Solomon** codes

Algebraic coding theory



k-symbol message



Encoding



Channel



Decoding



k-symbol message

Interleaved Reed-Solomon codes

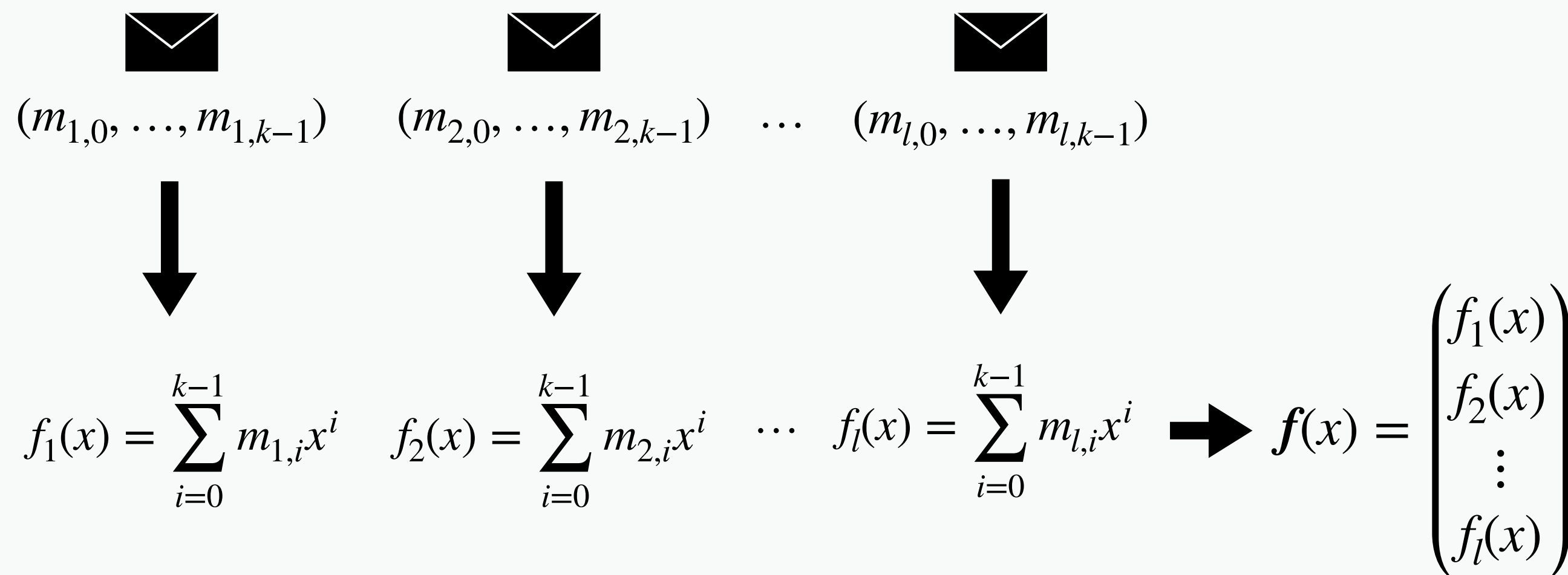
Interleaved Reed-Solomon Codes

Let $k \leq n \leq q$ and $\{\alpha_1, \dots, \alpha_n\}$ distinct *evaluation points*,

$$\mathcal{C}_{IRS}(n, k) := \{(f(\alpha_1), \dots, f(\alpha_n)) \mid f \in \mathbb{F}_q[x]^{l \times 1}, \deg(f) < k\}$$

The IRS code is an **MDS code**

the **minimum distance** is $d = n - k + 1$



Encoding

$$C = \begin{pmatrix} f_1(\alpha_1) & f_1(\alpha_2) & f_1(\alpha_3) & \dots & f_1(\alpha_n) \\ f_2(\alpha_1) & f_2(\alpha_2) & f_2(\alpha_3) & \dots & f_2(\alpha_n) \\ f_3(\alpha_1) & f_3(\alpha_2) & f_3(\alpha_3) & \dots & f_3(\alpha_n) \\ \vdots & \vdots & \vdots & \dots & \vdots \\ f_l(\alpha_1) & f_l(\alpha_2) & f_l(\alpha_3) & \dots & f_l(\alpha_n) \end{pmatrix}$$

Interleaved Reed-Solomon codeword

Interleaved Reed-Solomon codes

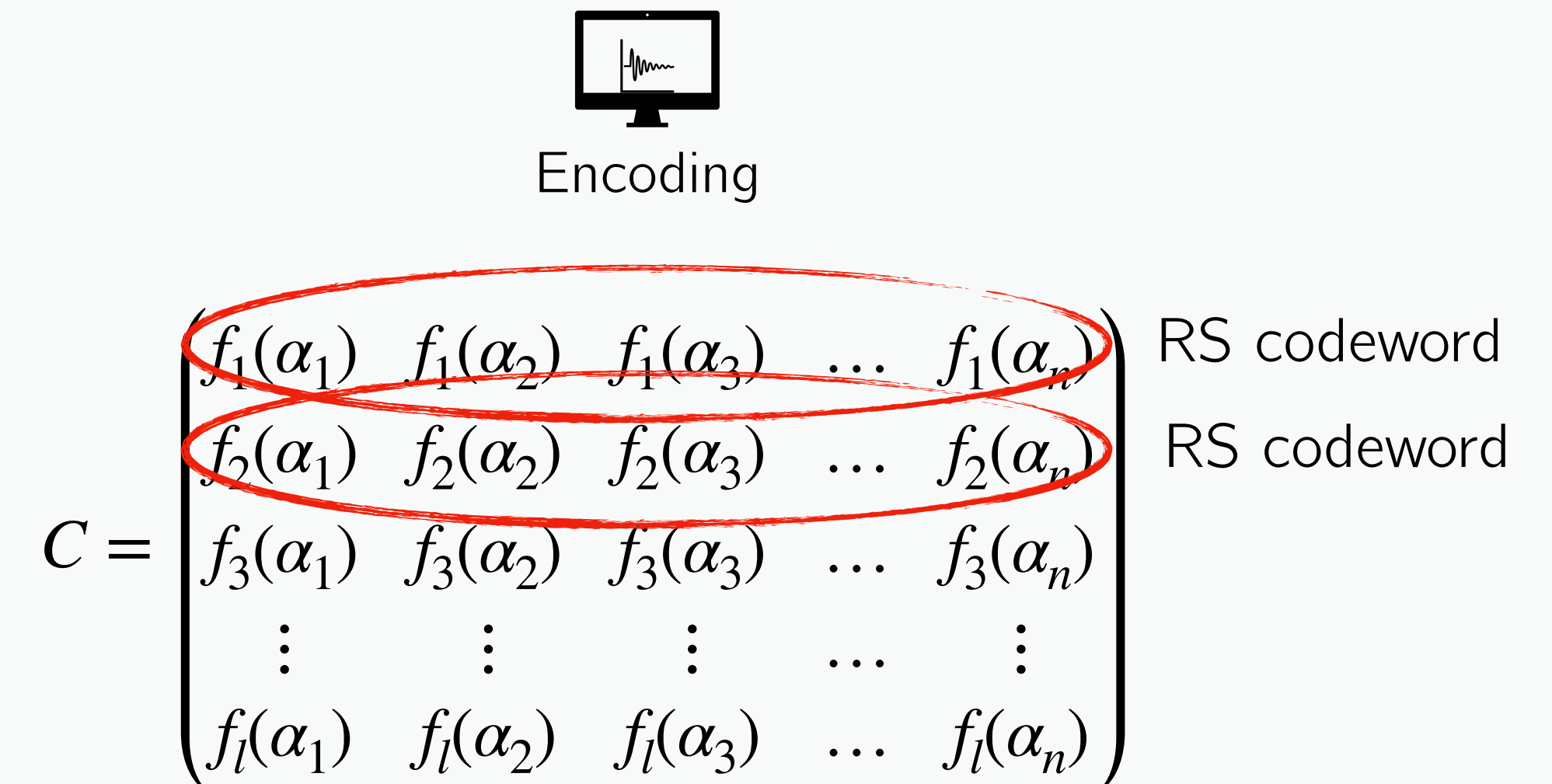
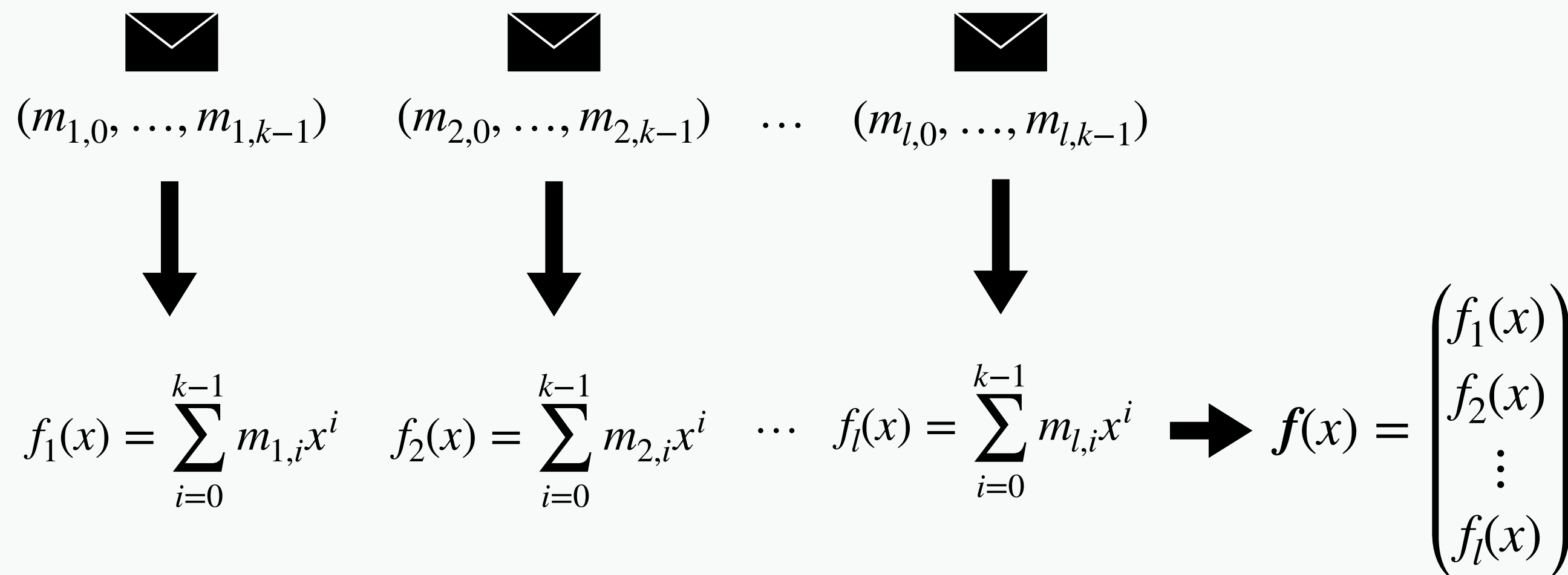
Interleaved Reed-Solomon Codes

Let $k \leq n \leq q$ and $\{\alpha_1, \dots, \alpha_n\}$ distinct *evaluation points*,

$$\mathcal{C}_{IRS}(n, k) := \{(f(\alpha_1), \dots, f(\alpha_n)) \mid f \in \mathbb{F}_q[x]^{l \times 1}, \deg(f) < k\}$$

The IRS code is an **MDS code**

the **minimum distance** is $d = n - k + 1$



Interleaved Reed-Solomon codeword

Interleaved Reed-Solomon codes

Interleaved Reed-Solomon Codes

Let $k \leq n \leq q$ and $\{\alpha_1, \dots, \alpha_n\}$ distinct *evaluation points*,

$$\mathcal{C}_{IRS}(n, k) := \{(f(\alpha_1), \dots, f(\alpha_n)) \mid f \in \mathbb{F}_q[x]^{l \times 1}, \deg(f) < k\}$$



Encoding

$$C = \begin{pmatrix} f_1(\alpha_1) & f_1(\alpha_2) & f_1(\alpha_3) & \dots & f_1(\alpha_n) \\ f_2(\alpha_1) & f_2(\alpha_2) & f_2(\alpha_3) & \dots & f_2(\alpha_n) \\ f_3(\alpha_1) & f_3(\alpha_2) & f_3(\alpha_3) & \dots & f_3(\alpha_n) \\ \vdots & \vdots & \vdots & \dots & \vdots \\ f_l(\alpha_1) & f_l(\alpha_2) & f_l(\alpha_3) & \dots & f_l(\alpha_n) \end{pmatrix}$$



Channel

$$\begin{pmatrix} f_1(\alpha_1) \\ f_2(\alpha_1) \\ f_3(\alpha_1) \\ \vdots \\ f_l(\alpha_1) \end{pmatrix}$$

The IRS code is an **MDS code**

the **minimum distance** is $d = n - k + 1$

Interleaved Reed-Solomon codes

Interleaved Reed-Solomon Codes

Let $k \leq n \leq q$ and $\{\alpha_1, \dots, \alpha_n\}$ distinct *evaluation points*,

$$\mathcal{C}_{IRS}(n, k) := \{(f(\alpha_1), \dots, f(\alpha_n)) \mid f \in \mathbb{F}_q[x]^{l \times 1}, \deg(f) < k\}$$



Encoding

$$C = \begin{pmatrix} f_1(\alpha_1) & f_1(\alpha_2) & f_1(\alpha_3) & \dots & f_1(\alpha_n) \\ f_2(\alpha_1) & f_2(\alpha_2) & f_2(\alpha_3) & \dots & f_2(\alpha_n) \\ f_3(\alpha_1) & f_3(\alpha_2) & f_3(\alpha_3) & \dots & f_3(\alpha_n) \\ \vdots & \vdots & \vdots & \dots & \vdots \\ f_l(\alpha_1) & f_l(\alpha_2) & f_l(\alpha_3) & \dots & f_l(\alpha_n) \end{pmatrix}$$



Channel

$$\begin{pmatrix} f_1(\alpha_1) & f_1(\alpha_2) \\ f_2(\alpha_1) & f_2(\alpha_2) \\ f_3(\alpha_1) & f_3(\alpha_2) \\ \vdots & \vdots \\ f_l(\alpha_1) & f_l(\alpha_2) \end{pmatrix}$$

The IRS code is an **MDS code**

the **minimum distance** is $d = n - k + 1$

Interleaved Reed-Solomon codes

Interleaved Reed-Solomon Codes

Let $k \leq n \leq q$ and $\{\alpha_1, \dots, \alpha_n\}$ distinct *evaluation points*,

$$\mathcal{C}_{IRS}(n, k) := \{(f(\alpha_1), \dots, f(\alpha_n)) \mid f \in \mathbb{F}_q[x]^{l \times 1}, \deg(f) < k\}$$

The IRS code is an **MDS code**

the **minimum distance** is $d = n - k + 1$



Encoding

$$C = \begin{pmatrix} f_1(\alpha_1) & f_1(\alpha_2) & f_1(\alpha_3) & \dots & f_1(\alpha_n) \\ f_2(\alpha_1) & f_2(\alpha_2) & f_2(\alpha_3) & \dots & f_2(\alpha_n) \\ f_3(\alpha_1) & f_3(\alpha_2) & f_3(\alpha_3) & \dots & f_3(\alpha_n) \\ \vdots & \vdots & \vdots & \dots & \vdots \\ f_l(\alpha_1) & f_l(\alpha_2) & f_l(\alpha_3) & \dots & f_l(\alpha_n) \end{pmatrix}$$



Channel

$$\begin{pmatrix} f_1(\alpha_1) & f_1(\alpha_2) \\ f_2(\alpha_1) & f_2(\alpha_2) \\ f_3(\alpha_1) & f_3(\alpha_2) \\ \vdots & \vdots \\ f_l(\alpha_1) & f_l(\alpha_2) \end{pmatrix}$$

↓
burst error

Interleaved Reed-Solomon codes

Interleaved Reed-Solomon Codes

Let $k \leq n \leq q$ and $\{\alpha_1, \dots, \alpha_n\}$ distinct *evaluation points*,

$$\mathcal{C}_{IRS}(n, k) := \{(f(\alpha_1), \dots, f(\alpha_n)) \mid f \in \mathbb{F}_q[x]^{l \times 1}, \deg(f) < k\}$$

The IRS code is an **MDS code**

the **minimum distance** is $d = n - k + 1$



Encoding

$$C = \begin{pmatrix} f_1(\alpha_1) & f_1(\alpha_2) & f_1(\alpha_3) & \dots & f_1(\alpha_n) \\ f_2(\alpha_1) & f_2(\alpha_2) & f_2(\alpha_3) & \dots & f_2(\alpha_n) \\ f_3(\alpha_1) & f_3(\alpha_2) & f_3(\alpha_3) & \dots & f_3(\alpha_n) \\ \vdots & \vdots & \vdots & \dots & \vdots \\ f_l(\alpha_1) & f_l(\alpha_2) & f_l(\alpha_3) & \dots & f_l(\alpha_n) \end{pmatrix}$$



Channel

$$\begin{pmatrix} f_1(\alpha_1) & f_1(\alpha_2) & f_1(\alpha_3) & \dots & f_1(\alpha_n) \\ f_2(\alpha_1) & f_2(\alpha_2) & f_2(\alpha_3) & \dots & f_2(\alpha_n) \\ f_3(\alpha_1) & f_3(\alpha_2) & f_3(\alpha_3) & \dots & f_3(\alpha_n) \\ \vdots & \vdots & \vdots & \dots & \vdots \\ f_l(\alpha_1) & f_l(\alpha_2) & f_l(\alpha_3) & \dots & f_l(\alpha_n) \end{pmatrix}$$

↓
burst error

Interleaved Reed-Solomon codes

Interleaved Reed-Solomon Codes

Let $k \leq n \leq q$ and $\{\alpha_1, \dots, \alpha_n\}$ distinct *evaluation points*,

$$\mathcal{C}_{IRS}(n, k) := \{(f(\alpha_1), \dots, f(\alpha_n)) \mid f \in \mathbb{F}_q[x]^{l \times 1}, \deg(f) < k\}$$



Decoding

$$Y = \begin{pmatrix} f_1(\alpha_1) & f_1(\alpha_2) & f_1(\alpha_3) & \dots & f_1(\alpha_n) \\ f_2(\alpha_1) & f_2(\alpha_2) & f_2(\alpha_3) & \dots & f_2(\alpha_n) \\ f_3(\alpha_1) & f_3(\alpha_2) & f_3(\alpha_3) & \dots & f_3(\alpha_n) \\ \vdots & \vdots & \vdots & \dots & \vdots \\ f_l(\alpha_1) & f_l(\alpha_2) & f_l(\alpha_3) & \dots & f_l(\alpha_n) \end{pmatrix}$$

↓
burst error

The IRS code is an **MDS code**
the **minimum distance** is $d = n - k + 1$

Decoding IRS codes

Given the received matrix with columns $\mathbf{y}_1, \dots, \mathbf{y}_n$

- $\mathbf{y}_j = f(\alpha_j)$ correct evaluations
- $\mathbf{y}_j \neq f(\alpha_j)$ erroneous evaluations

the *degree bound* $k > \deg(f)$

and an *upper bound* τ on the number of errors.

GOAL: **reconstruct** $f(x)$

Interleaved Reed-Solomon codes

Interleaved Reed-Solomon Codes

Let $k \leq n \leq q$ and $\{\alpha_1, \dots, \alpha_n\}$ distinct *evaluation points*,

$$\mathcal{C}_{IRS}(n, k) := \{(f(\alpha_1), \dots, f(\alpha_n)) \mid f \in \mathbb{F}_q[x]^{l \times 1}, \deg(f) < k\}$$

The IRS code is an **MDS code**

the **minimum distance** is $d = n - k + 1$

Simultaneous Interpolation with Errors

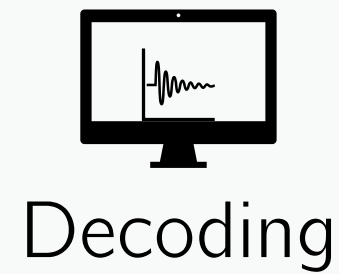
Given the received matrix with columns $\mathbf{y}_1, \dots, \mathbf{y}_n$

- $\mathbf{y}_j = f(\alpha_j)$ correct evaluations
- $\mathbf{y}_j \neq f(\alpha_j)$ erroneous evaluations

the *degree bound* $k > \deg(f)$

and an *upper bound* τ on the number of errors.

GOAL: **reconstruct** $f(x)$



Decoding

this is exactly our starting point!
Simultaneous Interpolation with Errors

$$Y = \begin{pmatrix} f_1(\alpha_1) & f_1(\alpha_2) & f_1(\alpha_3) & \dots & f_1(\alpha_n) \\ f_2(\alpha_1) & f_2(\alpha_2) & f_2(\alpha_3) & \dots & f_2(\alpha_n) \\ f_3(\alpha_1) & f_3(\alpha_2) & f_3(\alpha_3) & \dots & f_3(\alpha_n) \\ \vdots & \vdots & \vdots & \dots & \vdots \\ f_l(\alpha_1) & f_l(\alpha_2) & f_l(\alpha_3) & \dots & f_l(\alpha_n) \end{pmatrix}$$

burst error



Simultaneous Cauchy Interpolation

Decoding Interleaved Reed-Solomon codes

Simultaneous Interpolation with Errors

Given the received matrix with columns $\mathbf{y}_1, \dots, \mathbf{y}_n$

- $\mathbf{y}_j = f(\alpha_j)$ correct evaluations
- $\mathbf{y}_j \neq f(\alpha_j)$ erroneous evaluations

the degree bound $k > \deg(f)$

and an upper bound τ on the number of errors.

GOAL: reconstruct $f(x)$



Simultaneous Cauchy Interpolation

Given the vectors $\mathbf{y}_1, \dots, \mathbf{y}_n$

and the degree bounds $\tau + k, \tau + 1$

GOAL: find $(\underbrace{\varphi_1(x), \dots, \varphi_l(x)}_{\varphi(x)}, \psi(x))$ s.t.

- $\varphi_i(\alpha_j) = y_{i,j} \psi(\alpha_j)$
- $\deg(\varphi_i) < \tau + k$
- $\deg(\psi) < \tau + 1$



Simultaneous Rational Function Reconstruction

Decoding Interleaved Reed-Solomon codes

Simultaneous Interpolation with Errors

Given the received matrix with columns $\mathbf{y}_1, \dots, \mathbf{y}_n$

- $\mathbf{y}_j = f(\alpha_j)$ correct evaluations
- $\mathbf{y}_j \neq f(\alpha_j)$ erroneous evaluations

the degree bound $k > \deg(f)$

and an upper bound τ on the number of errors.

GOAL: reconstruct $f(x)$



Simultaneous Cauchy Interpolation

Given the vectors $\mathbf{y}_1, \dots, \mathbf{y}_n$

and the degree bounds $\tau + k, \tau + 1$

GOAL: find $(\underbrace{\varphi_1(x), \dots, \varphi_l(x)}_{\varphi(x)}, \psi(x))$ s.t.

- $\Lambda(\alpha_j) f_i(\alpha_j) = y_{i,j} \Lambda(\alpha_j)$
- $\deg(\Lambda f) < \tau + k$
- $\deg(\Lambda) < \tau + 1$

$(\Lambda(x)f(x), \Lambda(x))$

solution

$$\Lambda(x) = \prod_{\alpha_j \text{ erroneous}} (x - \alpha_j)$$

Error Locator Polynomial

roots = erroneous evaluation points

$$\deg(\Lambda) = \text{nb errors}$$

Decoding Interleaved Reed-Solomon codes

Simultaneous Interpolation with Errors

Given the received matrix with columns $\mathbf{y}_1, \dots, \mathbf{y}_n$

- $\mathbf{y}_j = f(\alpha_j)$ correct evaluations
- $\mathbf{y}_j \neq f(\alpha_j)$ erroneous evaluations

the degree bound $k > \deg(f)$

and an upper bound τ on the number of errors.

GOAL: reconstruct $f(x)$



Simultaneous Cauchy Interpolation

Given the vectors $\mathbf{y}_1, \dots, \mathbf{y}_n$

and the degree bounds $\tau + k, \tau + 1$

GOAL: find $(\underbrace{\varphi_1(x), \dots, \varphi_l(x)}_{\varphi(x)}, \psi(x))$ s.t.

- $\varphi_i(\alpha_j) = y_{i,j} \psi(\alpha_j)$
- $\deg(\varphi_i) < \tau + k$
- $\deg(\psi) < \tau + 1$

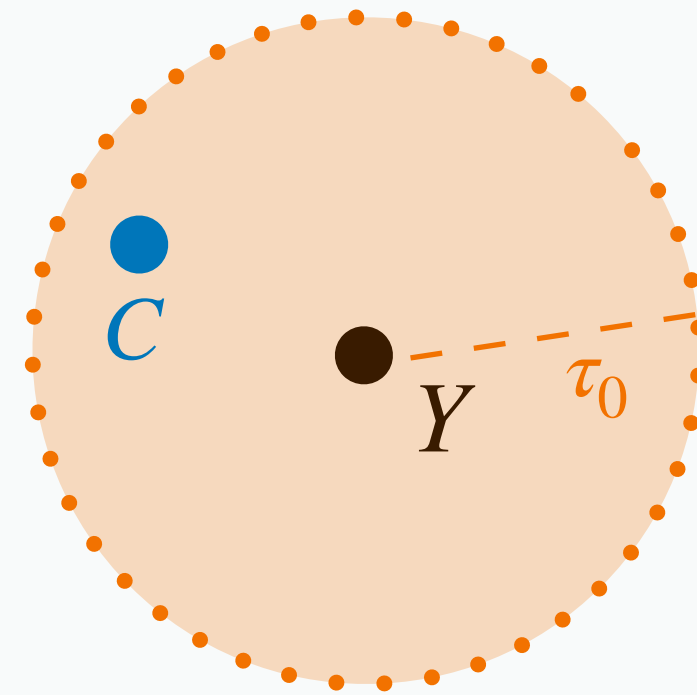


uniqueness

$(\Lambda(x)f(x), \Lambda(x))$

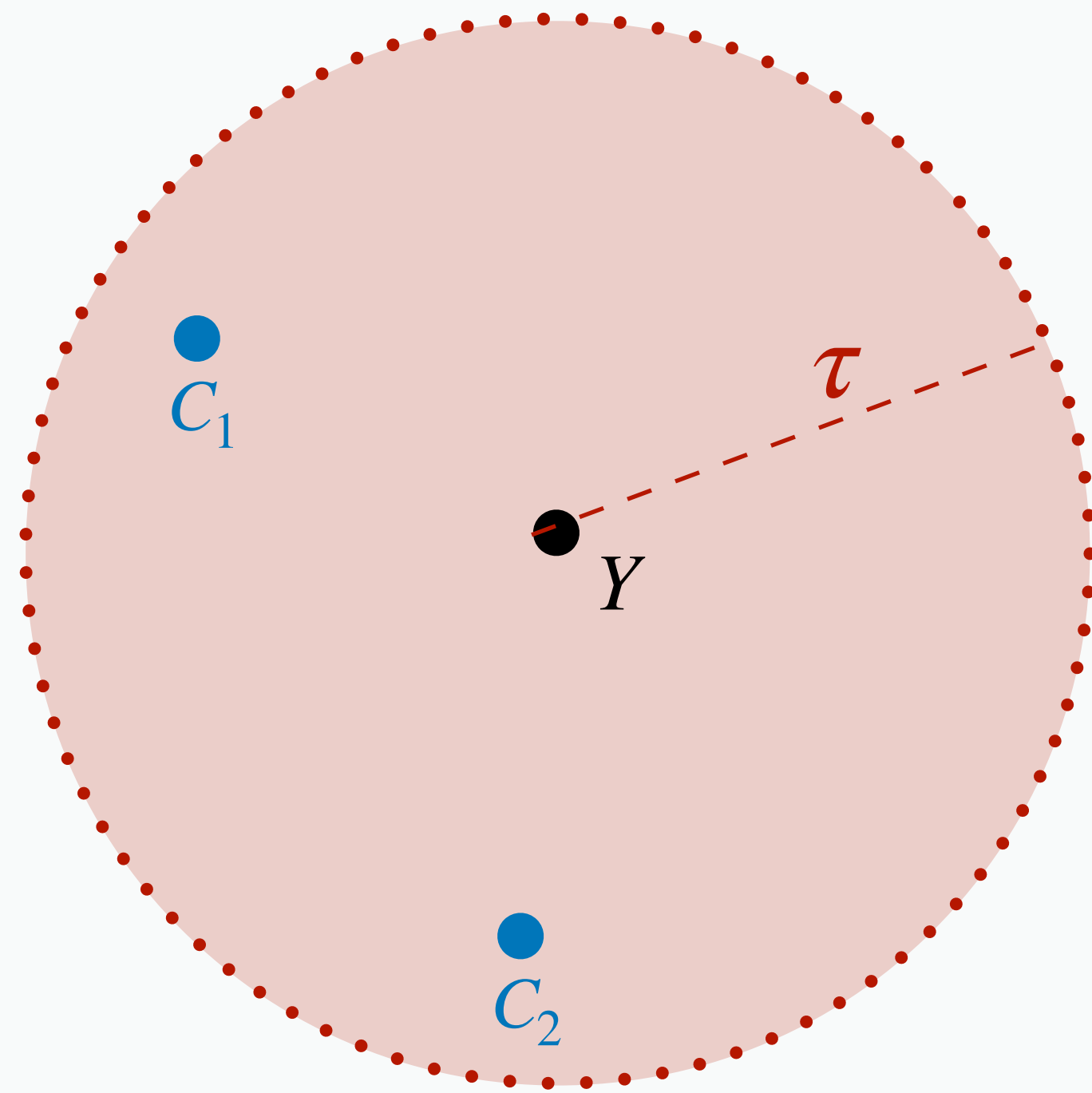


unique decoding



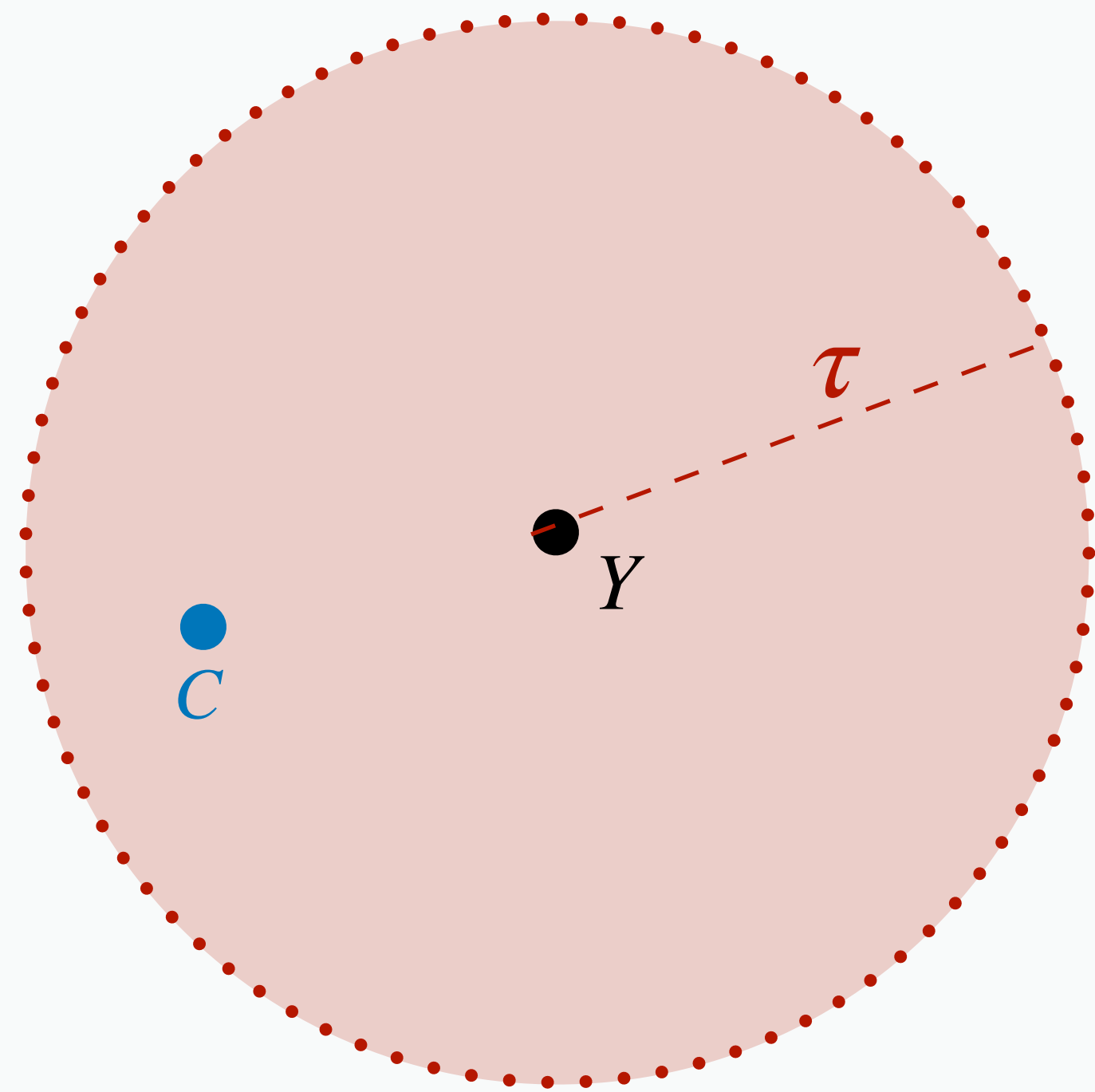
1. Unique decoding Capability

$$\text{nb errors} \leq \frac{n-k}{2} = \frac{d-1}{2} := \tau_0 \longrightarrow \text{unique decoding}$$



1. Unique decoding Capability

$$\text{nb errors} \leq \frac{n-k}{2} = \frac{d-1}{2} := \tau_0 \longrightarrow \text{unique decoding}$$



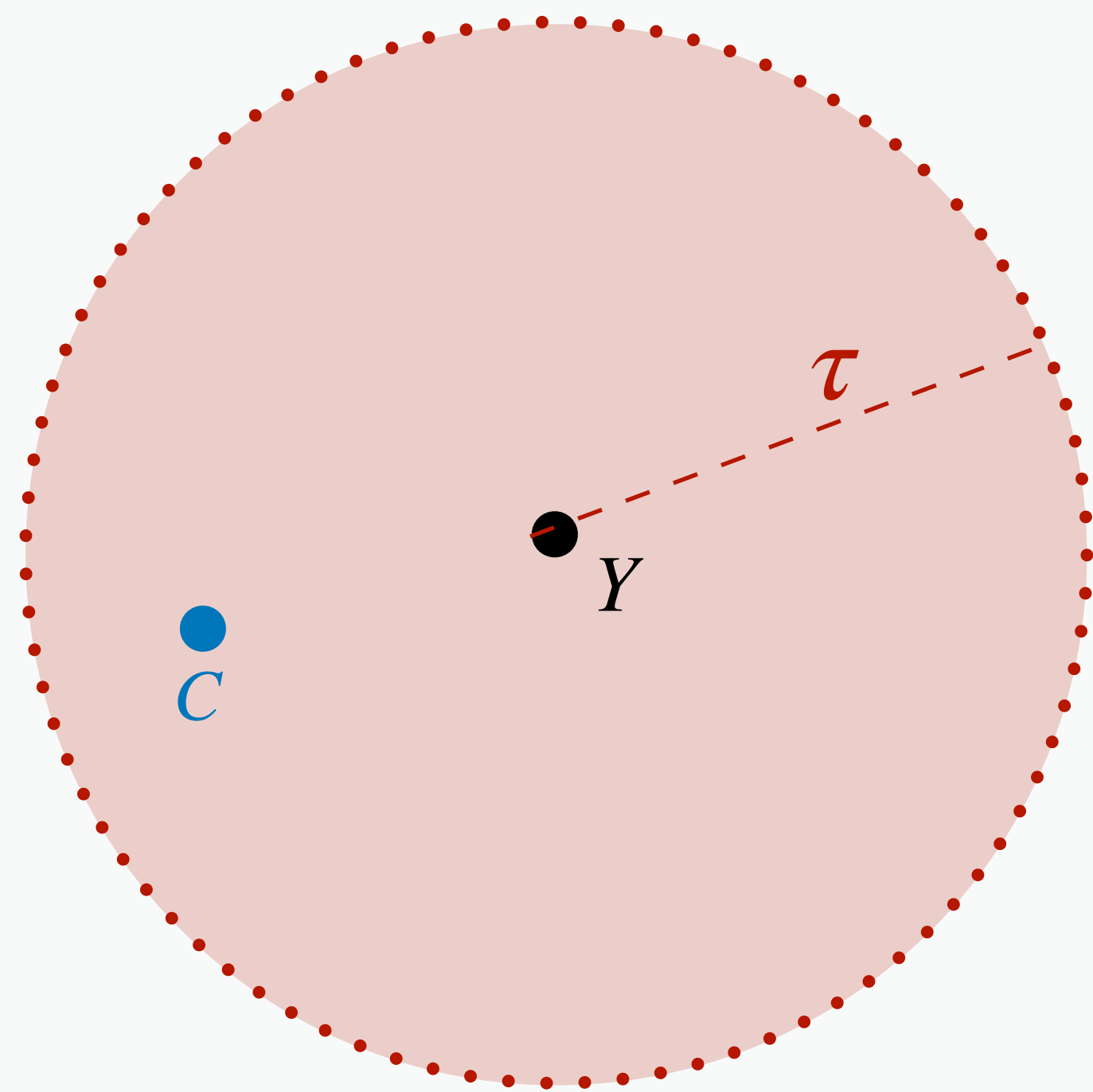
1. Unique decoding Capability

$$\text{nb errors} \leq \frac{n-k}{2} = \frac{d-1}{2} := \tau_0 \longrightarrow \text{unique decoding}$$

2. [BLEICHENBACHER, KIAYIAS, YUNG, 2003]

$$\text{nb errors} \leq \frac{l(n-k)}{l+1} =: \tau_{IRS} \longrightarrow \begin{array}{l} \text{uniqueness} \\ \text{for almost all} \\ \text{error patterns} \end{array}$$

The proportion of error patterns leading to non-uniqueness $\leq \text{nb errors}/q$



1. Unique decoding Capability

$$\text{nb errors} \leq \frac{n-k}{2} = \frac{d-1}{2} := \tau_0 \longrightarrow \text{unique decoding}$$

2. [BLEICHENBACHER, KIAYIAS, YUNG, 2003]

$$\text{nb errors} \leq \frac{l(n-k)}{l+1} =: \tau_{IRS} \longrightarrow \begin{array}{l} \text{uniqueness} \\ \text{for almost all} \\ \text{error patterns} \end{array}$$

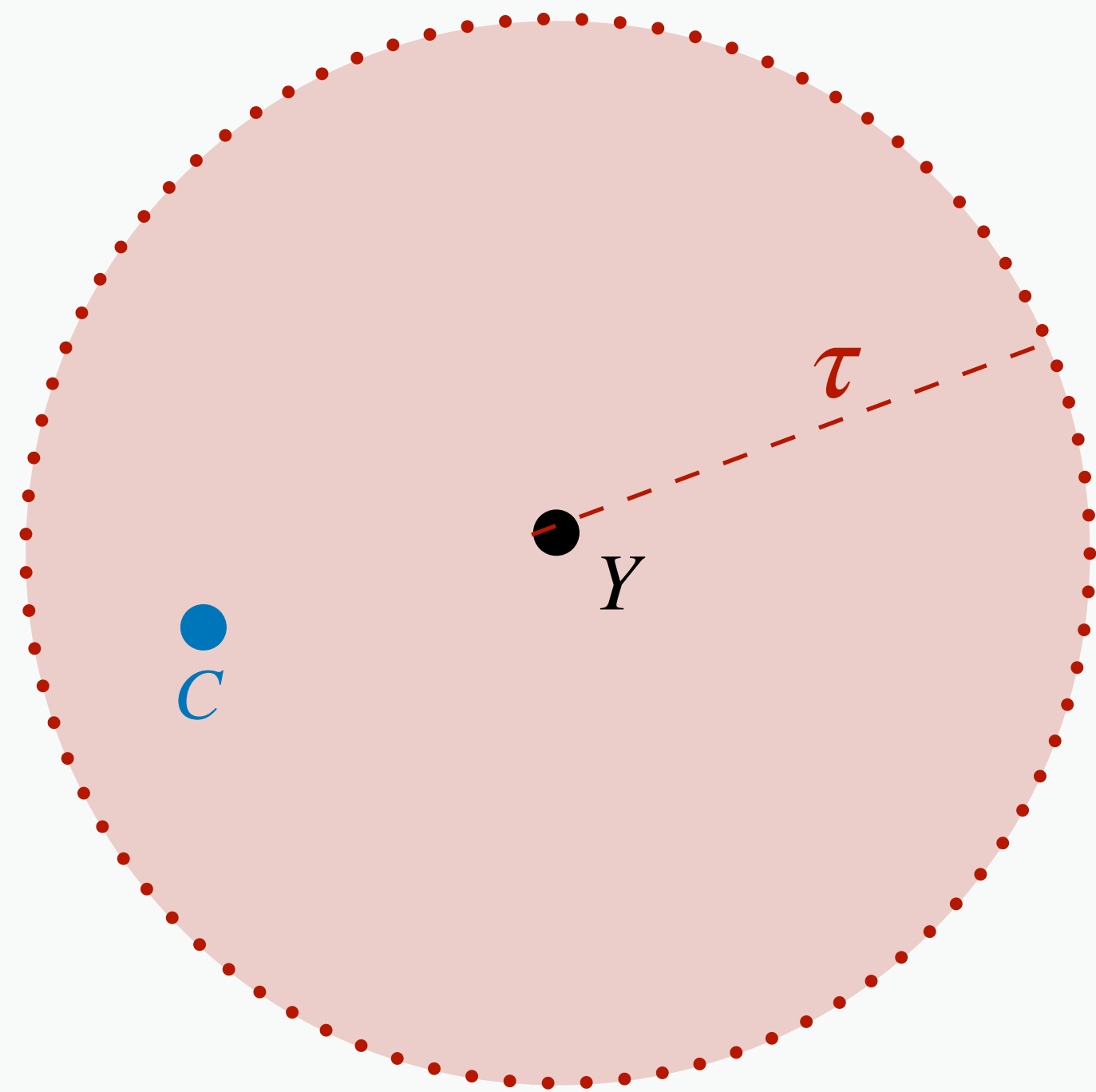
does not depend on errors, $\mathcal{O}(1/q)$

[BROWN, MINDER, SHOKROLLAHI, 2004]

[SCHIMDT, SIDORENKO, BOSSERT, 2009]

[PUCHINGER, ROSENKILDE, 2017]

The proportion of error patterns leading to non-uniqueness $\leq \text{nb errors}/q$



1. Unique decoding Capability

$$\text{nb errors} \leq \frac{n-k}{2} = \frac{d-1}{2} := \tau_0 \longrightarrow \text{unique decoding}$$

2. [BLEICHENBACHER, KIAYIAS, YUNG, 2003]

$$\text{nb errors} \leq \frac{l(n-k)}{l+1} =: \tau_{IRS} \longrightarrow \begin{array}{l} \text{uniqueness} \\ \text{for almost all} \\ \text{error patterns} \end{array}$$

The proportion of error patterns leading to non-uniqueness $\leq \text{nb errors}/q$

Simultaneous Cauchy Interpolation

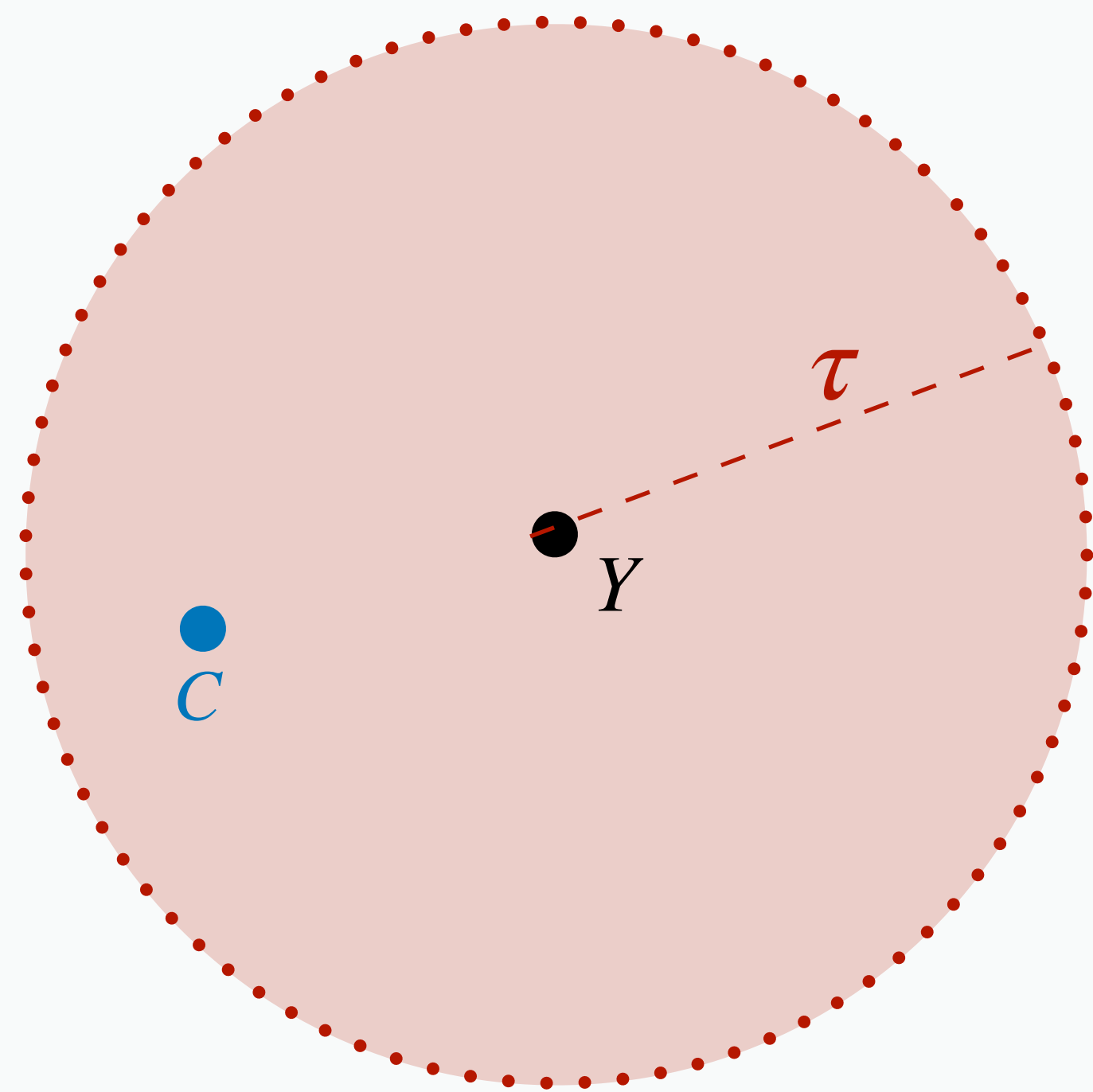
Given the vectors $\mathbf{y}_1, \dots, \mathbf{y}_n$
and the degree bounds $\tau + k, \tau + 1$

GOAL: **find** $(\underbrace{\varphi_1(x), \dots, \varphi_l(x)}_{\varphi(x)}, \psi(x))$ s.t.

- $\varphi_i(\alpha_j) = y_{i,j} \psi(\alpha_j)$
- $\deg(\varphi_i) < \tau + k$
- $\deg(\psi) < \tau + 1$

1. Cauchy Interpolation component-wise (RFR)

$$n = \tau + k + \tau \iff \text{nb errors} \leq \frac{n-k}{2} =: \tau_0 \longrightarrow \text{uniqueness}$$



Simultaneous Cauchy Interpolation

Given the vectors $\mathbf{y}_1, \dots, \mathbf{y}_n$
and the degree bounds $\tau + k, \tau + 1$

GOAL: **find** $(\underbrace{\varphi_1(x), \dots, \varphi_l(x)}_{\varphi(x)}, \psi(x))$ s.t.

- $\varphi_i(\alpha_j) = y_{i,j} \psi(\alpha_j)$
- $\deg(\varphi_i) < \tau + k$
- $\deg(\psi) < \tau + 1$

1. Unique decoding Capability

$$\text{nb errors} \leq \frac{n - k}{2} = \frac{d - 1}{2} := \tau_0 \longrightarrow \text{unique decoding}$$

2. [BLEICHENBACHER, KIAYIAS, YUNG, 2003]

$$\text{nb errors} \leq \frac{l(n - k)}{l + 1} =: \tau_{IRS} \longrightarrow \begin{array}{l} \text{uniqueness} \\ \text{for almost all} \\ \text{error patterns} \end{array}$$

1. Cauchy Interpolation component-wise (RFR)

$$n = \tau + k + \tau \iff \text{nb errors} \leq \frac{n - k}{2} =: \tau_0 \longrightarrow \text{uniqueness}$$

2. Common denominator feature

$$n = k + \left\lceil \frac{\tau}{l} \right\rceil + \tau \iff \text{nb errors} \leq \frac{l(n - k)}{l + 1} =: \tau_{IRS} \longrightarrow \begin{array}{l} \text{uniqueness} \\ \text{for almost all} \\ \text{error patterns} \end{array}$$

The proportion of error patterns leading to non-uniqueness $\leq \text{nb errors}/q$

Simultaneous Cauchy Interpolation with Errors

Simultaneous Interpolation with Errors

Reconstruct a **vector of polynomials**
by its evaluations, some erroneous



decoding IRS codes



Simultaneous Cauchy Interpolation

Simultaneous Cauchy Interpolation with Errors

Simultaneous Interpolation with Errors

Reconstruct a **vector of polynomials**
by its evaluations, some erroneous




↓
decoding IRS codes

Simultaneous Cauchy Interpolation with Errors




Reconstruct a **vector of rational functions**
by its evaluations, some erroneous

⚡ **Simultaneous Cauchy Interpolation**

Number of Evaluations - Outline of this work

		uniqueness	uniqueness <i>almost always</i>
no-errors	$(\mathbf{v}(x), d(x))$	$L = N + D - 1$ Cauchy Interpolation	 $L = N + \left\lceil \frac{(D-1)}{n} \right\rceil$
	$A(x) \frac{\mathbf{v}(x)}{d(x)} = \mathbf{b}(x)$	$L = \max\{\deg(A) + N, \deg(\mathbf{b}) + D\}$ [CABAY, 1971]	
with errors	$(\mathbf{v}(x), d(x))$	d constant ($D = 0$)	$L = N - 1 + 2\tau$ Unique Decoding Capability Theorem (IRS codes) [BLEICHENBACHER, KIAYIAS, YUNG, 2003]
			[BOYER, KALTOFEN, 2014]  [GUERRINI, LEBRETON, Z., 2019]
	$A(x) \frac{\mathbf{v}(x)}{d(x)} = \mathbf{b}(x)$	$D > 0$	[KALTOFEN, PERNET, STORJOHANN, WADDEL, 2017]  [GUERRINI, LEBRETON, Z., 2020]

Number of Evaluations - Outline of this work

		uniqueness	uniqueness <i>almost always</i>
no-errors	$(\mathbf{v}(x), d(x))$	$L = N + D - 1$ Cauchy Interpolation	 $L = N + \left\lceil \frac{(D-1)}{n} \right\rceil$
	$A(x) \frac{\mathbf{v}(x)}{d(x)} = \mathbf{b}(x)$	$L = \max\{\deg(A) + N, \deg(\mathbf{b}) + D\}$ [CABAY, 1971]	
with errors	$(\mathbf{v}(x), d(x))$	d constant ($D = 0$)	$L = N - 1 + 2\tau$ Unique Decoding Capability Theorem (IRS codes) [BLEICHENBACHER, KIAYIAS, YUNG, 2003]
		$D > 0$	 [GUERRINI, LEBRETON, Z., 2019]
	$A(x) \frac{\mathbf{v}(x)}{d(x)} = \mathbf{b}(x)$	$D > 0$	[KALTOFEN, PERNET, STORJOHANN, WADDEL, 2017]  [GUERRINI, LEBRETON, Z., 2020]

Simultaneous Cauchy Interpolation with Errors

Simultaneous Cauchy Interpolation with Errors

Given $\mathbf{y}_1, \dots, \mathbf{y}_L$

- $\mathbf{y}_j = \frac{\mathbf{v}(\alpha_j)}{d(\alpha_j)}$ correct evaluations
- $\mathbf{y}_j \neq \frac{\mathbf{v}(\alpha_j)}{d(\alpha_j)}$ erroneous evaluations

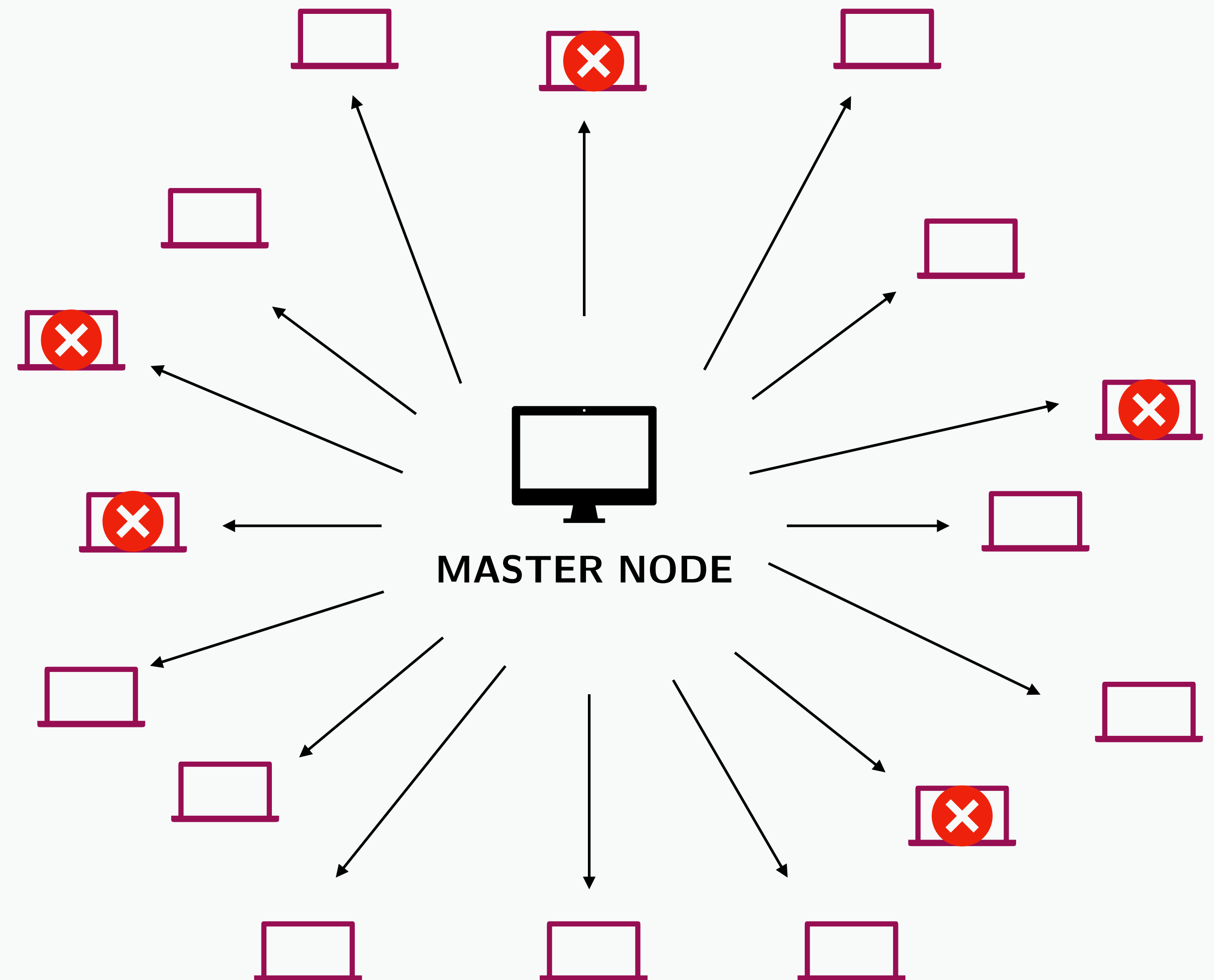
the degree bounds $N > \deg(\mathbf{v})$, $D > \deg(d)$
and an upper bound τ on the number of errors.

GOAL: reconstruct $(\mathbf{v}(x), d(x)) \rightarrow \mathbf{y}(x)$



Simultaneous Cauchy Interpolation

Simultaneous Rational Function Reconstruction



Simultaneous Cauchy Interpolation with Errors

Simultaneous Cauchy Interpolation with Errors

Given $\mathbf{y}_1, \dots, \mathbf{y}_L$

- $\mathbf{y}_j = \frac{\mathbf{v}(\alpha_j)}{d(\alpha_j)}$ correct evaluations
- $\mathbf{y}_j \neq \frac{\mathbf{v}(\alpha_j)}{d(\alpha_j)}$ erroneous evaluations

the degree bounds $N > \deg(\mathbf{v})$, $D > \deg(d)$
and an upper bound τ on the number of errors.

GOAL: reconstruct $(\mathbf{v}(x), d(x)) \rightarrow \mathbf{y}(x)$



Simultaneous Cauchy Interpolation

Given the vectors $\mathbf{y}_1, \dots, \mathbf{y}_L$
and the degree bounds $N + \tau$, $D + \tau$

GOAL: find $(\underbrace{\varphi_1(x), \dots, \varphi_n(x)}_{\varphi(x)}, \psi(x))$ s.t.

- $\varphi_i(\alpha_j) = y_{i,j} \psi(\alpha_j)$
- $\deg(\varphi_i) < N + \tau$
- $\deg(\psi) < D + \tau$

$(\Lambda(x)\mathbf{v}(x), \Lambda(x)d(x))$

solution

$$\Lambda(x) = \prod_{\alpha_j \text{ erroneous}} (x - \alpha_j)$$

Error Locator Polynomial

roots = erroneous evaluation points

$\deg(\Lambda) = \text{nb errors}$

Simultaneous Cauchy Interpolation with Errors

Simultaneous Cauchy Interpolation with Errors

Given $\mathbf{y}_1, \dots, \mathbf{y}_L$

- $\mathbf{y}_j = \frac{\mathbf{v}(\alpha_j)}{d(\alpha_j)}$ correct evaluations
- $\mathbf{y}_j \neq \frac{\mathbf{v}(\alpha_j)}{d(\alpha_j)}$ erroneous evaluations

the degree bounds $N > \deg(\mathbf{v})$, $D > \deg(d)$
and an upper bound τ on the number of errors.

GOAL: reconstruct $(\mathbf{v}(x), d(x)) \rightarrow \mathbf{y}(x)$



Simultaneous Cauchy Interpolation

Given the vectors $\mathbf{y}_1, \dots, \mathbf{y}_L$
and the degree bounds $N + \tau$, $D + \tau$

GOAL: find $(\underbrace{\varphi_1(x), \dots, \varphi_n(x)}_{\varphi(x)}, \psi(x))$ s.t.

- $\Lambda(\alpha_j) \mathbf{v}_i(\alpha_j) = y_{i,j} \Lambda(\alpha_j)$
- $\deg(\Lambda \mathbf{v}_i) < N + \tau$
- $\deg(\Lambda d) < D + \tau$

$(\Lambda(x)\mathbf{v}(x), \Lambda(x)d(x))$

solution

$$\Lambda(x) = \prod_{\alpha_j \text{ erroneous}} (x - \alpha_j)$$

Error Locator Polynomial

roots = erroneous evaluation points

$\deg(\Lambda) = \text{nb errors}$

Simultaneous Cauchy Interpolation with Errors

Simultaneous Cauchy Interpolation with Errors

Given $\mathbf{y}_1, \dots, \mathbf{y}_L$

- $\mathbf{y}_j = \frac{\mathbf{v}(\alpha_j)}{d(\alpha_j)}$ correct evaluations
- $\mathbf{y}_j \neq \frac{\mathbf{v}(\alpha_j)}{d(\alpha_j)}$ erroneous evaluations

the degree bounds $N > \deg(\mathbf{v})$, $D > \deg(d)$
and an upper bound τ on the number of errors.

GOAL: reconstruct $(\mathbf{v}(x), d(x)) \rightarrow \mathbf{y}(x)$



Simultaneous Cauchy Interpolation

Given the vectors $\mathbf{y}_1, \dots, \mathbf{y}_L$
and the degree bounds $N + \tau$, $D + \tau$

GOAL: find $(\underbrace{\varphi_1(x), \dots, \varphi_n(x)}_{\varphi(x)}, \psi(x))$ s.t.

- $\varphi_i(\alpha_j) = y_{i,j} \psi(\alpha_j)$
- $\deg(\varphi_i) < N + \tau$
- $\deg(\psi) < D + \tau$



uniqueness

$(\Lambda(x)\mathbf{v}(x), \Lambda(x)d(x))$

Simultaneous Cauchy Interpolation with Errors

Simultaneous Cauchy Interpolation

Given the vectors $\mathbf{y}_1, \dots, \mathbf{y}_L$

and the degree bounds $N + \tau, D + \tau$

GOAL: find $(\underbrace{\varphi_1(x), \dots, \varphi_n(x)}_{\varphi(x)}, \psi(x))$ s.t.

- $\varphi_i(\alpha_j) = y_{i,j} \psi(\alpha_j)$
- $\deg(\varphi_i) < N + \tau$
- $\deg(\psi) < D + \tau$

generalizing and re-elaborating the
result of [BLEICHENBACHER, KIAYIAS, YUNG, 2003]
for IRS codes

1. Cauchy Interpolation component-wise

$$L = N + D + 2\tau - 1 \longrightarrow \text{uniqueness}$$

[BOYER, KALTOFEN, 2014]

2. If we want to recover a solution of a PLS

$$L = \max\{\deg(A) + N, \deg(\mathbf{b}) + D\} + 2\tau \longrightarrow \text{uniqueness}$$

[CABAY, 1971] \longrightarrow [KALTOFEN, PERNET, STORJOHANN, WADDEL, 2017]

$$L = N + D - 1 + \left\lceil \frac{\tau}{n} \right\rceil + \tau \longrightarrow \text{uniqueness}$$

for almost all error patterns




If we want to recover a solution of a PLS

$$L = \max\{\deg(A) + N, \deg(\mathbf{b}) + D\} + \left\lceil \frac{\tau}{n} \right\rceil + \tau \longrightarrow \text{uniqueness}$$




for almost all error patterns

The proportion of error patterns leading to non-uniqueness $\leq (D + \tau)/q$

Number of Evaluations - Outline of this work

			uniqueness	uniqueness <i>almost always</i>
no-errors	$(\mathbf{v}(x), d(x))$		$L = N + D - 1$ Cauchy Interpolation	 $L = N + \left\lceil \frac{(D-1)}{n} \right\rceil$
	$A(x) \frac{\mathbf{v}(x)}{d(x)} = \mathbf{b}(x)$		$L = \max\{\deg(A) + N, \deg(\mathbf{b}) + D\}$ [CABAY, 1971]	
with errors	$(\mathbf{v}(x), d(x))$	d constant ($D = 0$)	$L = N - 1 + 2\tau$ Unique Decoding Capability Theorem (IRS codes)	$L = N - 1 + \left\lceil \frac{\tau}{n} \right\rceil + \tau$ [BLEICHENBACHER, KIAYIAS, YUNG, 2003]
		$D > 0$	$L = N + D - 1 + 2\tau$ [BOYER, KALTOFEN, 2014]	 $L = N + D - 1 + \left\lceil \frac{\tau}{n} \right\rceil + \tau$
	$A(x) \frac{\mathbf{v}(x)}{d(x)} = \mathbf{b}(x)$		$L = \max\{\deg(A) + N, \deg(\mathbf{b}) + D\} + 2\tau$ [KALTOFEN, PERNET, STORJOHANN, WADDEL, 2017]	 $L = \max\{\deg(A) + N, \deg(\mathbf{b}) + D\} + \left\lceil \frac{\tau}{n} \right\rceil + \tau$

Number of Evaluations - Outline of this work

		uniqueness	uniqueness <i>almost always</i>
no-errors	$(v(x), d(x))$	$L = N + D - 1$ Cauchy Interpolation	 $L = N + \left\lceil \frac{(D-1)}{n} \right\rceil$
	$A(x) \frac{v(x)}{d(x)} = b(x)$	$L = \max\{\deg(A) + N, \deg(b) + D\}$ [CABAY, 1971]	
with errors	$(v(x), d(x))$	d constant ($D = 0$)	$L = N - 1 + 2\tau$ Unique Decoding Capability Theorem (IRS codes) [BLEICHENBACHER, KIAYIAS, YUNG, 2003]
		$D > 0$	 $L = N + D - 1 + \left\lceil \frac{\tau}{n} \right\rceil + \tau$
	$A(x) \frac{v(x)}{d(x)} = b(x)$	$L = \max\{\deg(A) + N, \deg(b) + D\} + 2\tau$ [KALTOFEN, PERNET, STORJOHANN, WADDEL, 2017]	 $L = \max\{\deg(A) + N, \deg(b) + D\} + \left\lceil \frac{\tau}{n} \right\rceil + \tau$

Early Termination

$$\bar{L} = \min\{L_{BK}, L_{KPS}\}$$

- $L_{BK} = N + D + 2\tau - 1,$
- $L_{KPSW} = \max\{\deg(A) + N, \deg(\mathbf{b}) + D\} + 2\tau$

 \geq

$$L_{new} = \min\{L_{GLZ19}, L_{GLZ20}\}$$

- $L_{GLZ19} = N + D - 1 + \tau + \left\lceil \frac{\tau}{n} \right\rceil,$
- $L_{GLZ20} = \max\{\deg(A) + N, \deg(\mathbf{b}) + D\} + \tau + \left\lceil \frac{\tau}{n} \right\rceil$

All these bounds depend on

- $N > \deg(\mathbf{v})$
- $D > \deg(d)$
- $\tau \geq \text{nb errors } |E|$

we don't know these quantities

If N, D, τ are too big compared to $\deg(\mathbf{v}), \deg(d), |E| \longrightarrow \bar{L}, L_{new}$ too big compared to the number we really need

EARLY TERMINATION TECHNIQUE [KALTOFEN, PERNET, STORJOHANN, WADDELL, 2017]

GOAL: decrease the number of evaluation points without knowing the real degrees

Early termination strategy

$\deg(\mathbf{v})$
 $\deg(d)$
nb errors

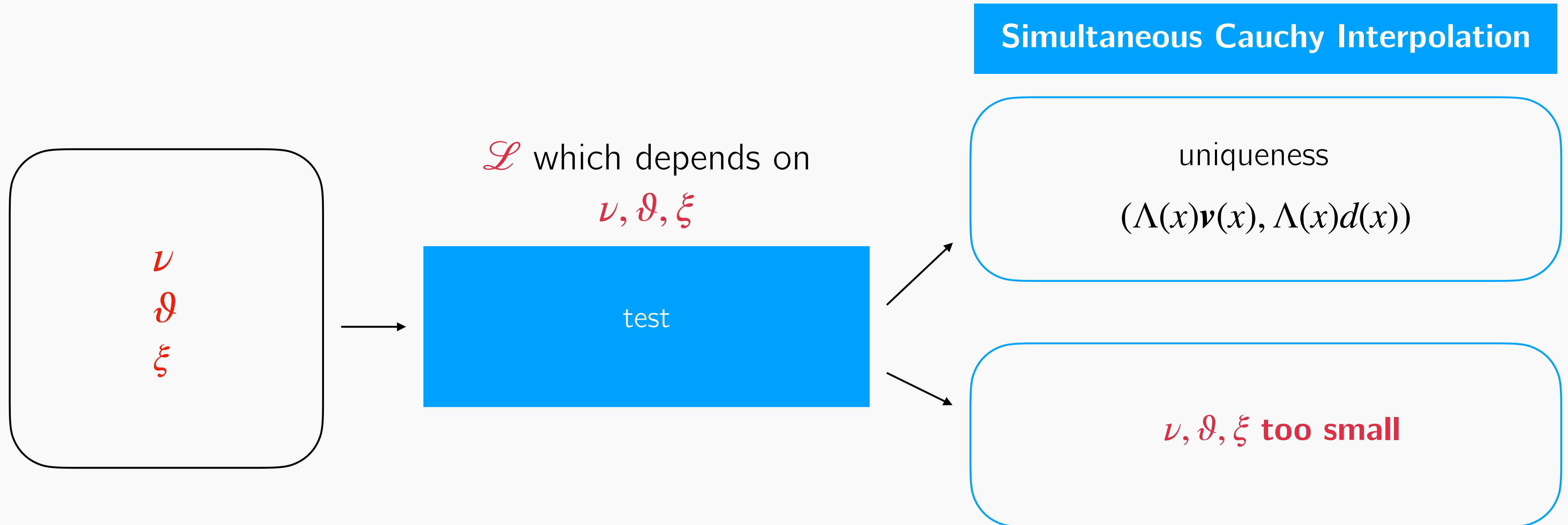
\mathcal{L} which depends on
 $\deg(\mathbf{v}), \deg(d), \mathbf{nb errors}$



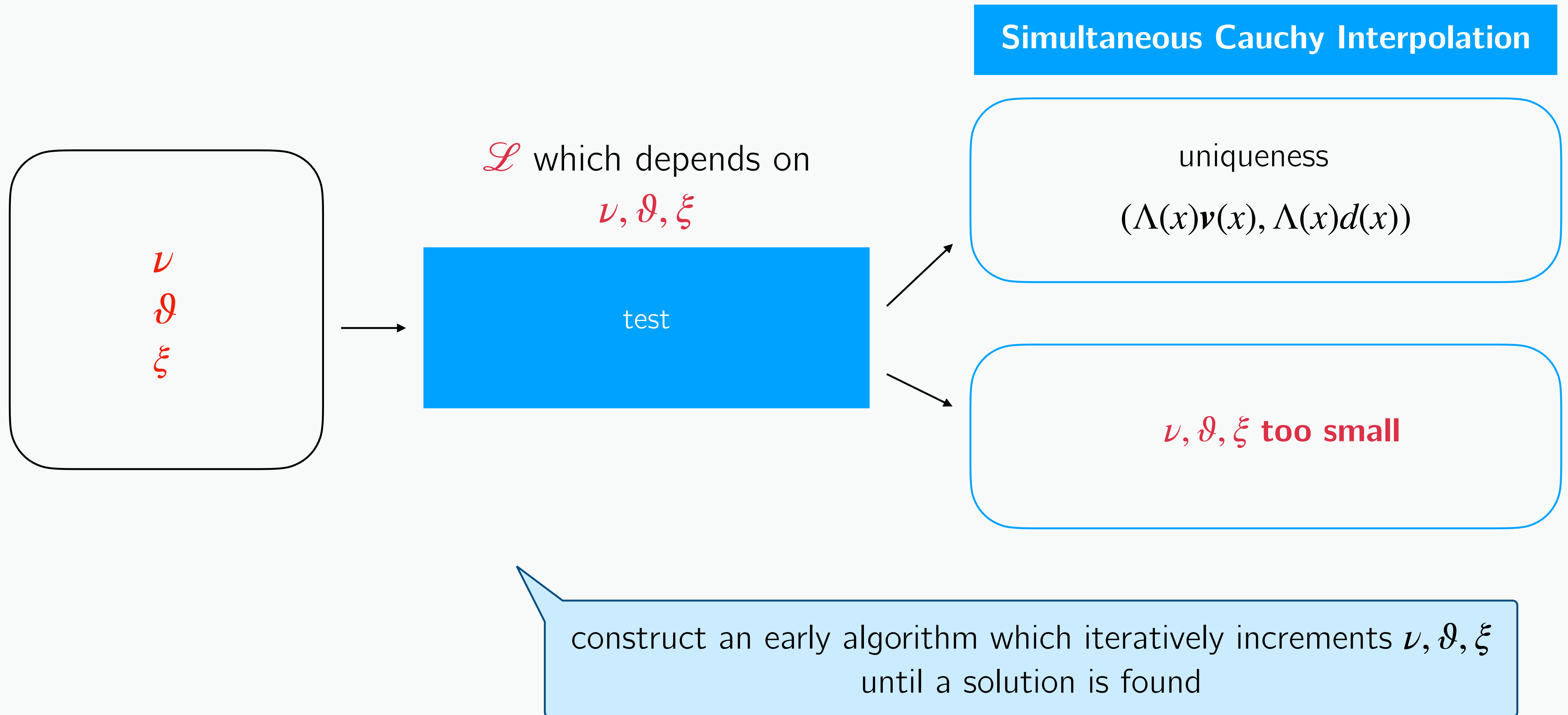
Simultaneous Cauchy Interpolation

uniqueness
 $(\Lambda(x)\mathbf{v}(x), \Lambda(x)d(x))$

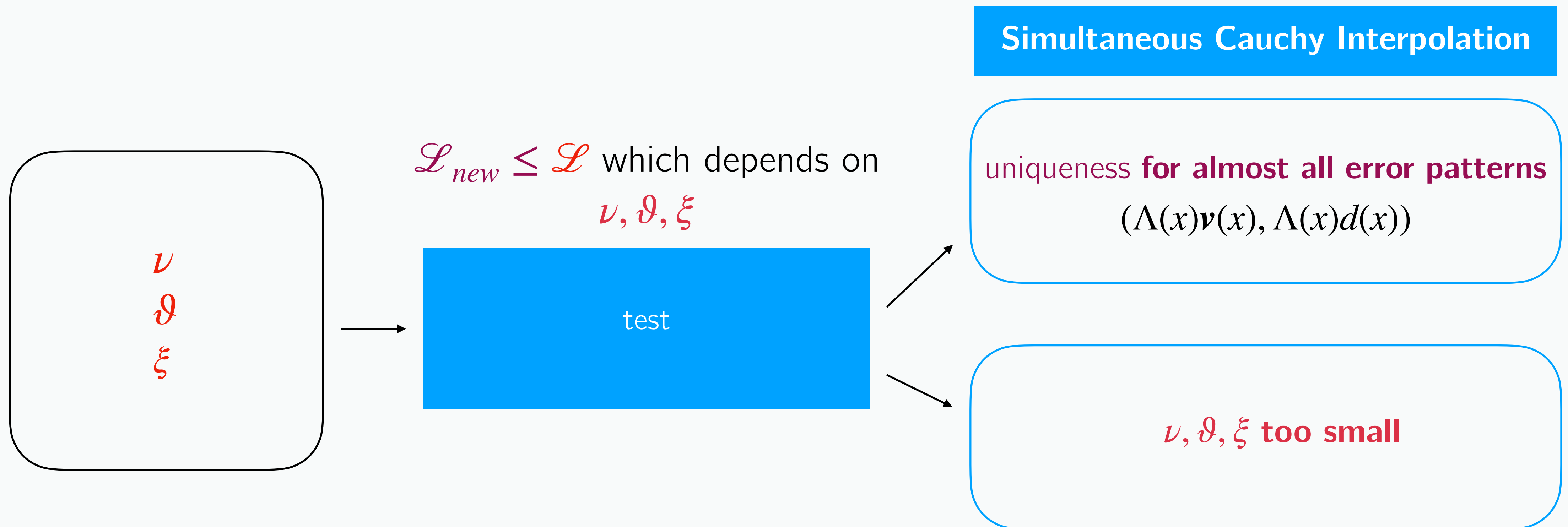
Early termination strategy



Early termination strategy



Early termination strategy





Conclusions & Open Problems



GT de l'équipe GRACE

December 1, 2020

Number of Evaluations - Outline of this work

		uniqueness	uniqueness <i>almost always</i>
no-errors	$(\mathbf{v}(x), d(x))$	$L = N + D - 1$ Cauchy Interpolation	 $L = N + \left\lceil \frac{(D-1)}{n} \right\rceil$
	$A(x) \frac{\mathbf{v}(x)}{d(x)} = \mathbf{b}(x)$	$L = \max\{\deg(A) + N, \deg(\mathbf{b}) + D\}$ [CABAY, 1971]	
with errors	$(\mathbf{v}(x), d(x))$	d constant ($D = 0$)	$L = N - 1 + 2\tau$ Unique Decoding Capability Theorem (IRS codes) [BLEICHENBACHER, KIAYIAS, YUNG, 2003]
		$D > 0$	$L = N + D - 1 + 2\tau$ [BOYER, KALTOFEN, 2014]
	$A(x) \frac{\mathbf{v}(x)}{d(x)} = \mathbf{b}(x)$	$L = \max\{\deg(A) + N, \deg(\mathbf{b}) + D\} + 2\tau$ [KALTOFEN, PERNET, STORJOHANN, WADDEL, 2017]	 $L = \max\{\deg(A) + N, \deg(\mathbf{b}) + D\} + \left\lceil \frac{\tau}{n} \right\rceil + \tau$

 **Early termination technique**

International conferences with proceedings

- 📄 ***Polynomial Linear System Solving with Errors by Simultaneous Polynomial Reconstruction of Interleaved Reed-Solomon codes.*** E. Guerrini, R. Lebreton, I. Zappatore. *In Proceedings of ISIT'19*, pages 1542-1546. IEEE, 2019
- 📄 ***On the Uniqueness of Rational Function Reconstruction.*** E. Guerrini, R. Lebreton, I. Zappatore. *In Proceedings of ISSAC'20*. ACM, 2020

Preprint, work in progress

- 📄 ***Enhancing Simultaneous Rational Function Recovery: adaptive error correction capability and new bounds for applications.*** E. Guerrini, R. Lebreton, I. Zappatore. *arXiv:2003.01793*

Improving previous results

- On the uniqueness of the simultaneous rational function reconstruction
- Failure probability of Simultaneous Cauchy Interpolation with Errors

Extending previous results

- Rational Function Codes
- Early termination techniques for decoding error correcting codes

Improving previous results

- On the uniqueness of the simultaneous rational function reconstruction
- Failure probability of Simultaneous Cauchy Interpolation with Errors

Extending previous results

- Rational Function Codes
- Early termination techniques for decoding error correcting codes

On the uniqueness of SRFR

Simultaneous Rational Function Reconstruction

Given two vector of polynomials $\mathbf{u}(x), \mathbf{a}(x)$
and the degree bounds N, D

GOAL: find $(v_1(x), \dots, v_n(x), d(x))$ s.t.

- $v_i(x) = \underbrace{v(x)}_{v(x)} d(x) \bmod a_i(x)$
- $\deg(v_i) < N$
- $\deg(d) < D$

$$\frac{v_i(x)}{d(x)} = u_i(x) \bmod a_i(x), \gcd(a_i, d) = 1$$

If $L = N + (D - 1)/n$, for almost all instances $u_i = v_i/d \implies$ uniqueness?

 **Theorem** [GUERRINI, LEBRETON, Z., 2020]

If $L = \deg(\mathbf{a}) = N + (D - 1)/n$, for almost all instances \implies uniqueness.

If $\mathbb{K} = \mathbb{F}_q$, the proportion of instances leading to non-uniqueness is $\leq (D - 1)/q$

Number of Evaluations - Outline of this work

		uniqueness	uniqueness <i>almost always</i>
no-errors	$\frac{v(x)}{d(x)}$	$L = N + D - 1$ Cauchy Interpolation	? $L = N + \left\lceil \frac{(D-1)}{n} \right\rceil$
	$A(x)\frac{v(x)}{d(x)} = b(x)$	$L = \max\{\deg(A) + N, \deg(b) + D\}$ [CABAY, 1971]	? $L = N + \left\lceil \frac{(D-1)}{n} \right\rceil$
with errors	$(v(x), d(x))$	d constant ($D = 0$)	$L = N - 1 + 2\tau$ Unique Decoding Capability Theorem (IRS codes) [BLEICHENBACHER, KIAYIAS, YUNG, 2003]
		$D > 0$	$L = N + D - 1 + 2\tau$ [BOYER, KALTOFEN, 2014]
	$A(x)\frac{v(x)}{d(x)} = b(x)$	$L = \max\{\deg(A) + N, \deg(b) + D\} + 2\tau$ [KALTOFEN, PERNET, STORJOHANN, WADDEL, 2017]	$L = \max\{\deg(A) + N, \deg(b) + D\} + \left\lceil \frac{\tau}{n} \right\rceil + \tau$

Simultaneous Cauchy Interpolation with Errors

Simultaneous Cauchy Interpolation

Given the vectors $\mathbf{y}_1, \dots, \mathbf{y}_L$

and the degree bounds $N + \tau, D + \tau$

GOAL: find $(\varphi_1(x), \dots, \varphi_n(x), \psi(x))$ s.t.

- $\varphi_i(\alpha_j) = y_{i,j} \psi(\alpha_j)$
- $\deg(\varphi_i) < N + \tau$
- $\deg(\psi) < D + \tau$



uniqueness

$(\Lambda(x)\mathbf{v}(x), \Lambda(x)d(x))$

1. Cauchy Interpolation component-wise (RFR)

$$L = N + D + 2\tau - 1 \longrightarrow \text{uniqueness}$$

[BOYER, KALTOFEN, 2014]

2. If we want to recover a solution of a PLS

$$L = \max\{\deg(A) + N, \deg(\mathbf{b}) + D\} + 2\tau \longrightarrow \text{uniqueness}$$

[CABAY, 1971] \longrightarrow [KALTOFEN, PERNET, STORJOHANN, WADDEL, 2014]

Common denominator constraint

$$L = N + \left\lceil \frac{D - 1 + \tau}{n} \right\rceil + \tau \longrightarrow \begin{array}{l} \text{uniqueness} \\ \text{for almost all error patterns} \\ \text{for almost all } \mathbf{v}/d ? \end{array}$$




Improving previous results

- On the uniqueness of the simultaneous rational function reconstruction
- Failure probability of Simultaneous Cauchy Interpolation with Errors

Extending previous results

- Rational Function Codes
- Early termination techniques for decoding error correcting codes

Number of Evaluations - Outline of this work

			uniqueness	uniqueness <i>almost always</i>
no-errors	$(\mathbf{v}(x), d(x))$		$L = N + D - 1$ Cauchy Interpolation	 $L = N + \left\lceil \frac{(D-1)}{n} \right\rceil$
	$A(x) \frac{\mathbf{v}(x)}{d(x)} = \mathbf{b}(x)$		$L = \max\{\deg(A) + N, \deg(\mathbf{b}) + D\}$ [CABAY, 1971]	?
with errors	$(\mathbf{v}(x), d(x))$	d constant ($D = 0$)	$L = N - 1 + 2\tau$ Unique Decoding Capability Theorem (IRS codes)	$L = N - 1 + \left\lceil \frac{\tau}{n} \right\rceil + \tau$ [BLEICHENBACHER, KIAYIAS, YUNG, 2003]
		$D > 0$	$L = N + D - 1 + 2\tau$ [BOYER, KALTOFEN, 2014]	 $L = N + D - 1 + \left\lceil \frac{\tau}{n} \right\rceil + \tau$
	$A(x) \frac{\mathbf{v}(x)}{d(x)} = \mathbf{b}(x)$		$L = \max\{\deg(A) + N, \deg(\mathbf{b}) + D\} + 2\tau$ [KALTOFEN, PERNET, STORJOHANN, WADDEL, 2017]	 $L = \max\{\deg(A) + N, \deg(\mathbf{b}) + D\} + \left\lceil \frac{\tau}{n} \right\rceil + \tau$

Simultaneous Cauchy Interpolation with Errors

Simultaneous Cauchy Interpolation

Given the vectors $\mathbf{y}_1, \dots, \mathbf{y}_L$

and the degree bounds $N + \tau, D + \tau$

GOAL: find $(\underbrace{\varphi_1(x), \dots, \varphi_n(x)}_{\varphi(x)}, \psi(x))$ s.t.

- $\varphi_i(\alpha_j) = y_{i,j} \psi(\alpha_j)$
- $\deg(\varphi_i) < N + \tau$
- $\deg(\psi) < D + \tau$

generalizing and re-elaborating the
result of [BLEICHENBACHER, KIAYIAS, YUNG, 2003]
for IRS codes

1. Cauchy Interpolation component-wise

$$L = N + D + 2\tau - 1 \longrightarrow \text{uniqueness}$$

[BOYER, KALTOFEN, 2014]

2. If we want to recover a solution of a PLS

$$L = \max\{\deg(A) + N, \deg(\mathbf{b}) + D\} + 2\tau \longrightarrow \text{uniqueness}$$

[CABAY, 1971] \longrightarrow [KALTOFEN, PERNET, STORJOHANN, WADDEL, 2017]

$$\text{📄 } L = N + D - 1 + \left\lceil \frac{\tau}{n} \right\rceil + \tau \longrightarrow \text{uniqueness for almost all error patterns}$$

- 📄 If we want to recover a solution of a PLS

$$L = \max\{\deg(A) + N, \deg(\mathbf{b}) + D\} + \left\lceil \frac{\tau}{n} \right\rceil + \tau \longrightarrow \text{uniqueness for almost all error patterns}$$

The proportion of error patterns leading to non-uniqueness $\leq (D + \tau)/q$

Simultaneous Cauchy Interpolation with Errors

Simultaneous Cauchy Interpolation

Given the vectors $\mathbf{y}_1, \dots, \mathbf{y}_L$

and the degree bounds $N + \tau, D + \tau$

GOAL: find $(\varphi_1(x), \dots, \varphi_n(x), \psi(x))$ s.t.

- $\varphi_i(\alpha_j) = y_{i,j} \psi(\alpha_j)$
- $\deg(\varphi_i) < N + \tau$
- $\deg(\psi) < D + \tau$

generalizing and re-elaborating the result of [BLEICHENBACHER, KIAYIAS, YUNG, 2003] for IRS codes

1. Cauchy Interpolation component-wise

$$L = N + D + 2\tau - 1 \longrightarrow \text{uniqueness}$$

[BOYER, KALTOFEN, 2014]

2. If we want to recover a solution of a PLS

$$L = \max\{\deg(A) + N, \deg(\mathbf{b}) + D\} + 2\tau \longrightarrow \text{uniqueness}$$

[CABAY, 1971] \longrightarrow [KALTOFEN, PERNET, STORJOHANN, WADDEL, 2017]

$$L = N + D + 2\tau - 1 + \lfloor \tau \rfloor \longrightarrow \text{uniqueness}$$

If we want to

$$L = \max\{\deg$$

can we improve this bound?
prove that it does not depend on errors as IRS?

[BROWN, MINDER, SHOKROLLAHI, 2004]

[SCHMIDT, SIDORENKO, BOSSERT, 2009]

[PUCHINGER, ROSENKILDE, 2017]

The proportion of error patterns leading to non-uniqueness $\leq (D + \tau)/q$

Improving previous results

- On the uniqueness of the simultaneous rational function reconstruction
- Failure probability of Simultaneous Cauchy Interpolation with Errors

Extending previous results

- Rational Function Codes
- Early termination techniques for decoding error correcting codes

Simultaneous Cauchy Interpolation with Errors

Simultaneous Interpolation with Errors

Reconstruct a **vector of polynomials**
by its evaluations, some erroneous

↓
decoding IRS codes

Simultaneous Cauchy Interpolation with Errors

Reconstruct a **vector of rational functions**
by its evaluations, some erroneous

↓
decoding specific
Interleaved Rational Function codes
[PERNET, 2017]



Simultaneous Cauchy Interpolation

Conclusions & Open Problems

Rational Function Code [PERNET, 2017]

generalization of Reed-Solomon codes,
non linear

Let $N, D \leq L \leq q$ and $\{\alpha_1, \dots, \alpha_L\}$ distinct *evaluation points*,

$$\mathcal{C}_{RF}(n, k) := \left\{ \left(\frac{v(\alpha_1)}{d(\alpha_1)}, \dots, \frac{v(\alpha_L)}{d(\alpha_L)} \right) \mid \frac{v}{d} \in \mathbb{F}_q(x), \deg(v) < N, \deg(d) < D, d(\alpha_j) \neq 0 \right\}$$

Interleaved Rational Function Code [PERNET, 2017]

the minimum distance $\geq L - (N + D + 2)$

Let $N, D \leq L \leq q$ and $\{\alpha_1, \dots, \alpha_L\}$ distinct *evaluation points*,

$$\mathcal{C}_{RF}(n, k) := \left\{ \left(\frac{v(\alpha_1)}{d(\alpha_1)}, \dots, \frac{v(\alpha_L)}{d(\alpha_L)} \right) \mid \frac{\mathbf{v}}{d} \in \mathbb{F}_q(x)^{n \times 1}, \deg(v) < N, \deg(d) < D, d_i(\alpha_j) \neq 0 \right\}$$

Improving previous results

- On the uniqueness of the simultaneous rational function reconstruction
- Failure probability of Simultaneous Cauchy Interpolation with Errors

Extending previous results

- Rational Function Codes \longrightarrow determine parameters and other applications
- Early termination techniques for decoding error correcting codes
[KHONJI, PERNET, ROCH, ROCHE, STALINSKI, 2010]



Thank you

GT de l'équipe GRACE

December 1, 2020