

# New Algorithms and Analyses for Sum- Preserving Encryption

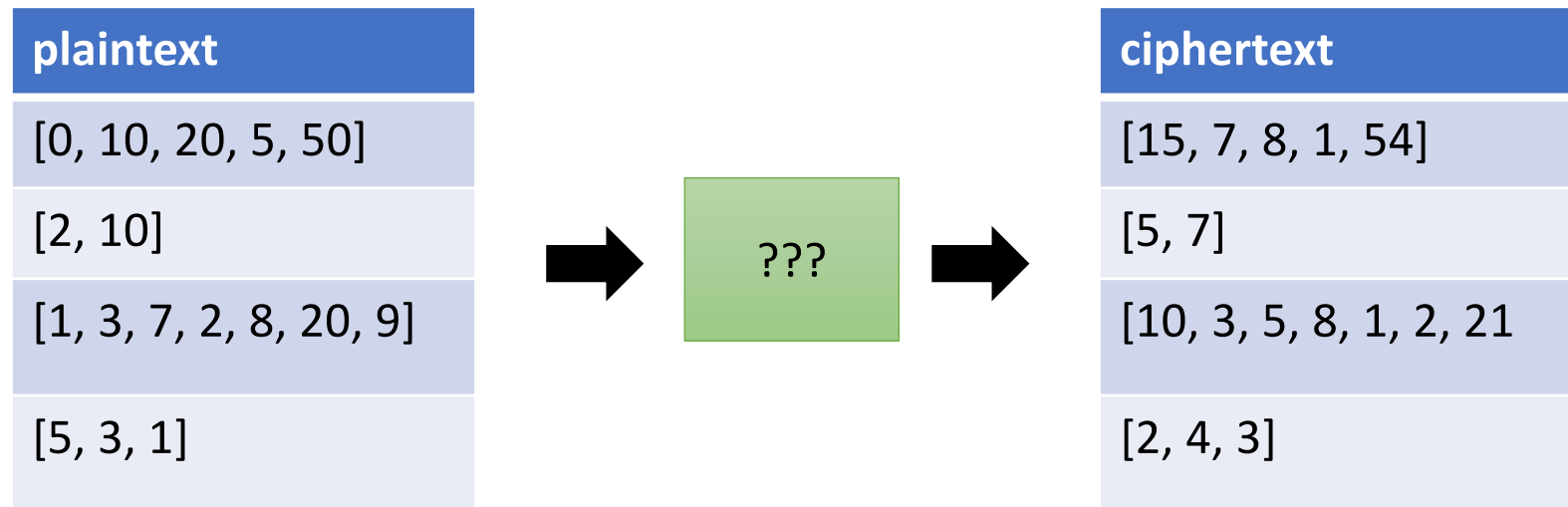
*Sarah Miracle* and Scott Yilek

University of St. Thomas

# Sum-Preserving Encryption

## Sum-Preserving Encryption schemes:

Encryption schemes in which ciphertexts and plaintexts are both integer vectors with the same sum.



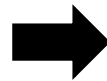
Vector components are typically bounded between 0 and  $d$ .

# Sum-Preserving Encryption

## Application: Thumbnail-preserving encryption

Image encryption where the thumbnail of an encrypted image matches the thumbnail of the unencrypted image.

- Divide the image into  $b \times b$  blocks of pixels
- Apply sum-preserving encryption to each block with component bound 255.

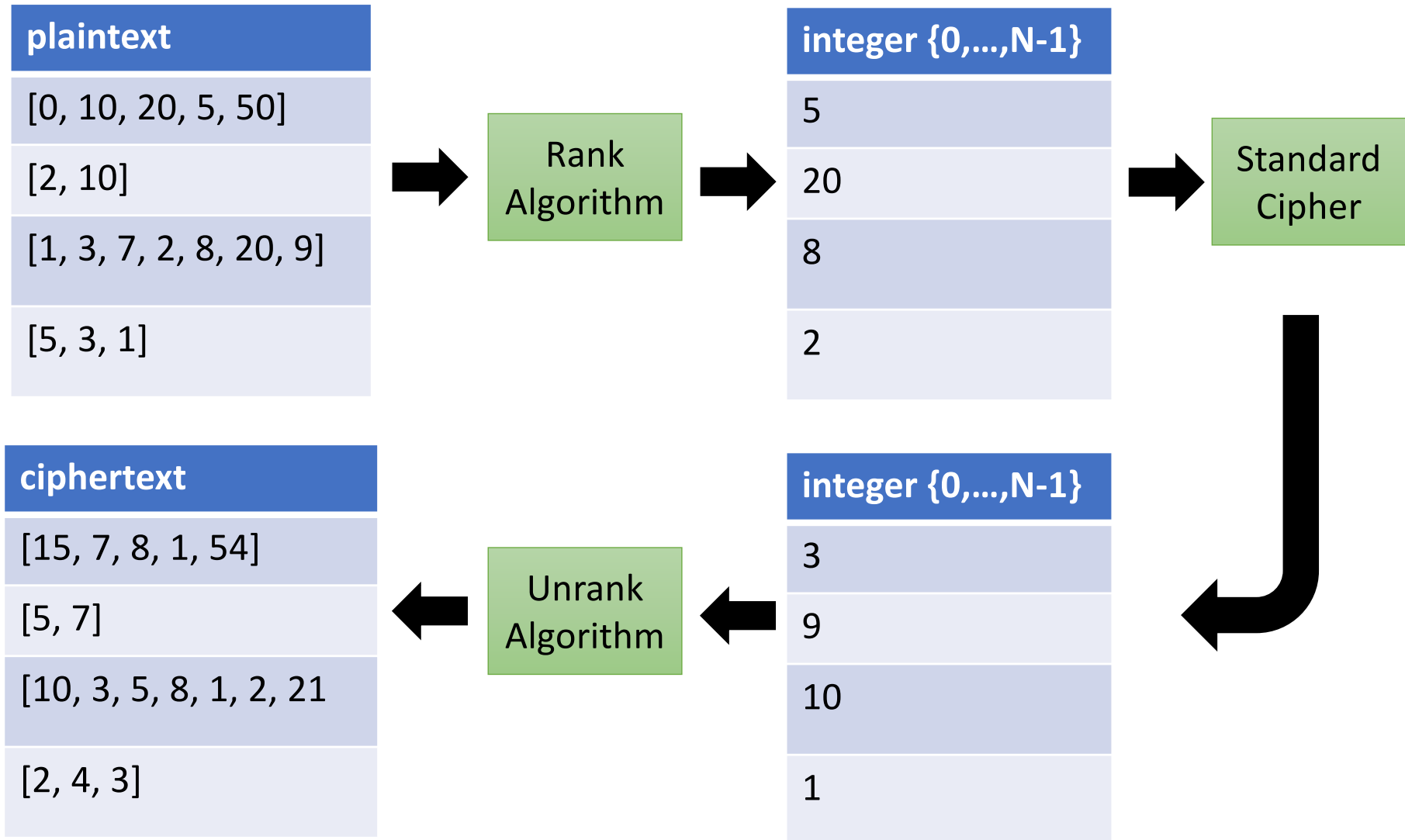


# Background

A special type of **format-preserving encryption**.

- Originally studied by Brightwell and Smith ['97]
- Formally defined and analyzed by Bellare, Ristenpart, Rogaway, and Stegers ['09]  
(use a rank-encipher-unrank construction)
- Widely studied and even standardized
- Tajik, Gunasekaran, Dutta, Ellis, Bobba, Rosulek, Wright, Feng, focus on problem of creating a thumbnail encryption scheme ['19]

# Rank-Encipher-Unrank



# Rank-Encipher-Unrank

Idea [Tajik et al.]:

[5, 3, 1]            11111011101

- Represent as a string of 1's and 0's
- The number of 1's is the same as the sum and 0's act as separators
- This is a regular language and thus we can use known techniques for ranking DFAs [see Bellare et al. '09]

However, high time complexity and impractical!

# Rank-Encipher-Unrank

Idea [Tajik et al.]:

[5, 3, 1]            11111011101

- Represent as a string of 1's and 0's
- The number of 1's is the same as the sum and 0's act as separators
- This is a regular language and thus we can use known techniques for ranking DFAs [see Bellare et al. '09]

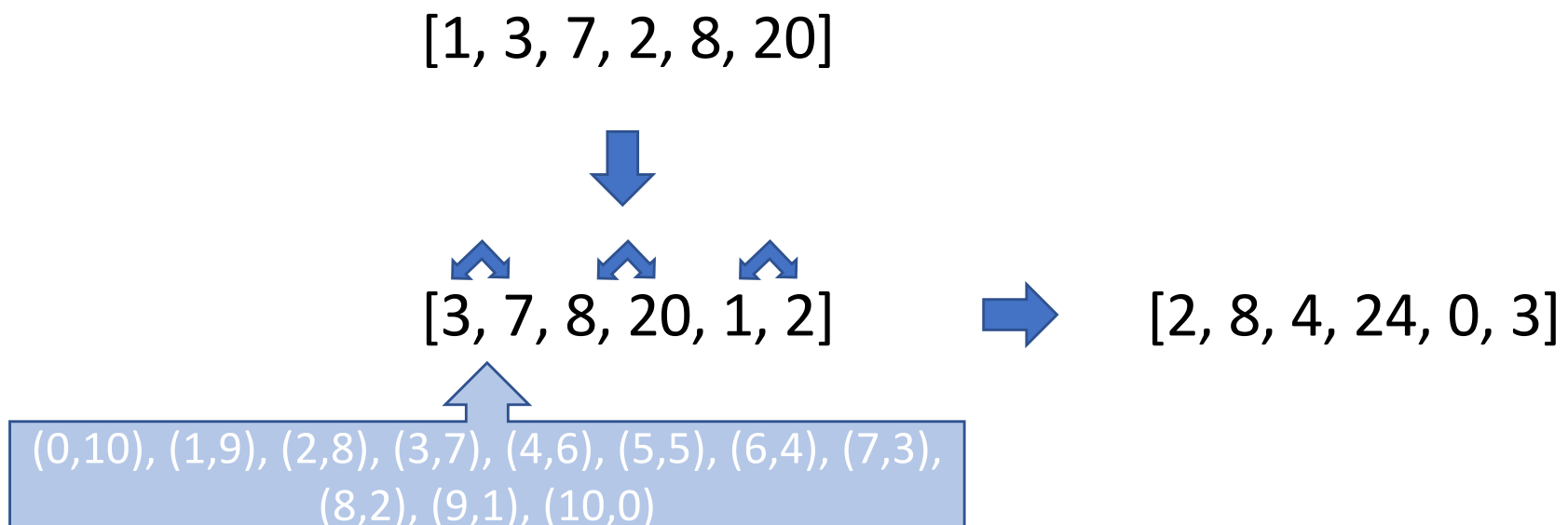
Ranking vectors of length 2 is simple & efficient!

# Tajik et al. Construction

## Sum-Preserving Shuffle Markov chain:

Repeat:

1. Choose a random shuffling on all points uniformly at random.
2. Pair adjacent points to create a perfect matching.
3. Independently for each matched pair select a pair of values u.a.r. from all valid choices that preserve the sum.





# Our Results

1. First proof bounding the mixing time of the Tajik et al. algorithm
2. Give practical rank and unrank algorithms for sum-preserving encryption
3. Create prototype implementations with performance comparisons

# Tajik et al. Construction

## Sum-Preserving Shuffle Markov chain:

Repeat:

1. Choose a random shuffling on all points uniformly at random.
2. Pair adjacent points to create a perfect matching.
3. Independently for each matched pair select a pair of values u.a.r. from all valid choices that preserve the sum.

How many times do you need to repeat?

# Tajik et al. Construction

## Sum-Preserving Shuffle Markov chain:

Repeat:

1. Choose a random shuffling on all points uniformly at random.
2. Pair adjacent points to create a perfect matching.
3. Independently for each matched pair select a pair of values u.a.r. from all valid choices that preserve the sum.

- Tajik et al. give heuristic arguments for what secure round choices might be but no proof.
- Test performance with 1000, 3000, and 5000 rounds.

# Mixing Time

Definition: The **total variation distance** is

$$\| P^t, \pi \| = \max_{x \in \Omega} \frac{1}{2} \sum_{y \in \Omega} |P^t(x, y) - \pi(y)|.$$

Definition: Given  $\varepsilon$ , the **mixing time** is

$$\tau(\varepsilon) = \min \{t: \|P^{t'}, \pi\| < \varepsilon, \quad \forall t' \geq t\}.$$

# Mixing Time Bound

## Sum-Preserving Shuffle Markov chain:

Repeat:

1. Choose a random shuffling on all points uniformly at random.
2. Pair adjacent points to create a perfect matching.
3. Independently for each matched pair select a pair of values u.a.r. from all valid choices that preserve the sum.

Let  $n$  be the vector length,  $d$  the component bound, and  $S$  the fixed sum. We show the mixing time satisfies

$$\tau(\epsilon) \leq n \ln(\min(dn, 2S)\epsilon^{-1})$$

# Proof Idea

Let  $n$  be the vector length,  $d$  the component bound, and  $S$  the fixed sum. We show the mixing time satisfies

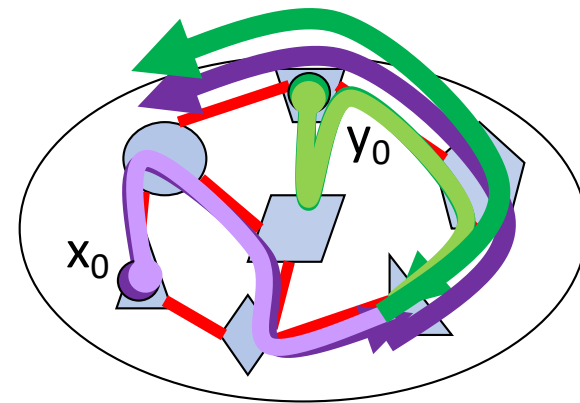
$$\tau(\epsilon) \leq n \ln(\min(dn, 2S)\epsilon^{-1})$$

Use **path coupling** due to Dyer, Greenhill ['98]

# Coupling

Simulate 2 processes:

- Start at any  $x_0$  and  $y_0$
- Couple moves, but each simulates the MC
- Once they agree, they move in sync  
( $x_t = y_t \rightarrow x_{t+1} = y_{t+1}$ )



Expected Coupling Time  $>$  Mixing time

Prove chains getting closer in expectation in each step



Manhattan distance

# Path Coupling

- **Coupling:** Show for all states  $x, y$ ,  
 $E[ \Delta (\text{dist}(x, y)) ] < 0$ .
- **Path coupling:** Show for all  $u, v$  s.t.  $u, v$  differ by 2 points,  
that  $E[ \Delta (\text{dist}(u, v)) ] < 0$ .

Consider a shortest path:

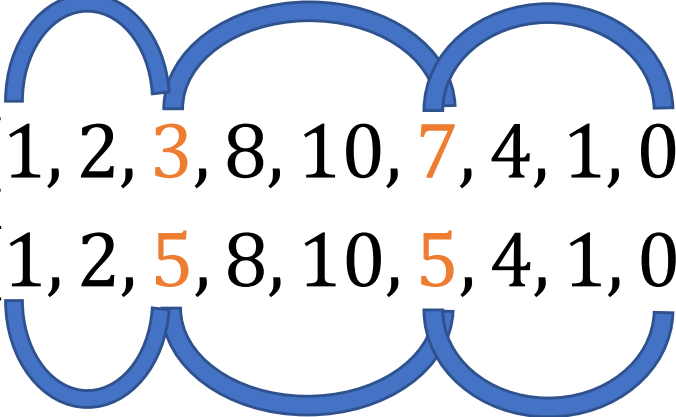
$x = z_0, z_1, z_2, \dots, z_r = y$ ,  $z_i, z_{i+1}$  differ by 2 points.

$$E[ \Delta (\text{dist}(x, y)) ] \leq \sum_i E[ \Delta (\text{dist}(z_i, z_{i+1})) ] \leq 0.$$



# Path Coupling

Consider 2 configuration that differ on exactly 2 points:


$$x = [1, 2, 3, 8, 10, 7, 4, 1, 0]$$
$$y = [1, 2, 5, 8, 10, 5, 4, 1, 0]$$

- If these 2 points are paired together the distance **decreases** to 0.
- Otherwise we show the distances **stays the same**.

# Rank-Encipher-Unrank

## Our Approach:

- Rank vectors directly using **lexicographical** order.
- Build on algorithms for unranking developed by Stein [‘20] for use in random sampling.
- Uses a dynamic programming approach and pre-computes a table  $C_d$  where  $C_d(n,S)$  stores the number of vectors of length  $n$  with sum  $S$  and component bound  $d$  (we instead store the cumulative sum)

Configuration	Rank	Configuration	Rank
(0,3,3)	0	(2,3,1)	5
(1,2,3)	1	(3,0,3)	6
(1,3,2)	2	(3,1,2)	7
(2,1,3)	3	(3,2,1)	8
(2,2,2)	4	(3,3,0)	9

Component bound = 3

# Recursive Block Order

To order  $x$  and  $y$ , divide each in half and compute the sum of each half.

$$S_{x_L} < S_{y_L} \implies x <_B y$$

$$S_{x_L} > S_{y_L} \implies x >_B y$$

$$S_{x_L} = S_{y_L} \text{ and } x_L <_B y_L \implies x <_B y$$

$$S_{x_L} = S_{y_L} \text{ and } x_L >_B y_L \implies x >_B y$$

$$S_{x_L} = S_{y_L} \text{ and } x_L = y_L \text{ and } x_R <_B y_R \implies x <_B y$$

$$S_{x_L} = S_{y_L} \text{ and } x_L = y_L \text{ and } x_R >_B y_R \implies x >_B y$$

# Recursive Block Order

$$S_{x_L} < S_{y_L} \implies x <_B y$$

$$S_{x_L} > S_{y_L} \implies x >_B y$$

$$S_{x_L} = S_{y_L} \text{ and } x_L <_B y_L \implies x <_B y$$

$$S_{x_L} = S_{y_L} \text{ and } x_L >_B y_L \implies x >_B y$$

$$S_{x_L} = S_{y_L} \text{ and } x_L = y_L \text{ and } x_R <_B x_L \implies x <_B y$$

$$S_{x_L} = S_{y_L} \text{ and } x_L = y_L \text{ and } x_R >_B x_L \implies x >_B y$$

Component bound = 3

Configuration	Rank	Configuration	Rank
(0,2,3,3)	0	(2,1,2,3)	7
(1,1,3,3)	1	(2,1,3,2)	8
(2,0,3,3)	2	(3,0,2,3)	9
(0,3,2,3)	3	(3,0,3,2)	10
(0,3,3,2)	4	(1,3,1,3)	11
(1,2,2,3)	5	(1,3,2,2)	12
(1,2,3,2)	6	(1,3,3,1)	13

# Recursive Block Order – Rank

$$S_{x_L} < S_{y_L}$$

$$\implies x <_B y$$

$$S_{x_L} > S_{y_L}$$

$$\implies x >_B y$$

$$S_{x_L} = S_{y_L} \text{ and } x_L <_B y_L$$

$$\implies x <_B y$$

$$S_{x_L} = S_{y_L} \text{ and } x_L >_B y_L$$

$$\implies x >_B y$$

$$S_{x_L} = S_{y_L} \text{ and } x_L = y_L \text{ and } x_R <_B x_L$$

$$\implies x <_B y$$

$$S_{x_L} = S_{y_L} \text{ and } x_L = y_L \text{ and } x_R >_B x_L$$

$$\implies x >_B y$$

$$\text{rank}_n(x) = \sum_{s=0}^{S_{x_L}-1} C(n/2, s) \cdot C(n/2, S - s) \\ + \text{rank}_{n/2}(x_L) \cdot C(n/2, S_{x_R}) \\ + \text{rank}_{n/2}(x_R)$$

We only need  $2 \log n$  rows!

# Filling the C table

- Dynamic Programming

$$C_d(n, S) = \begin{cases} 1 & \text{if } n = 1, S \leq d \\ 0 & \text{if } n = 1, S > d \\ 1 & \text{if } n = 0 \\ C_d(n-1, S) + C_d(n, S-1) & \text{if } n > 0, S \leq d \\ C_d(n-1, S) + C_d(n, S-1) - C_d(n-1, S-d-1) & \text{otherwise} \end{cases}$$

Start with 0     Start with number > 0

↓     ↓

- Generating Functions

Overcount by those that start with d

↑

$$C_d(n, S) = \sum_{k=0}^n (-1)^k \binom{n}{k} \binom{n + S - k(d + 1) - 1}{n - 1}$$

# Performance Tests and Results

Our implementation of Tajik et al. algorithm

<b>Application</b>	50 rounds	500 rounds	1000 rounds
10x10 image block ( $n = 100, d = 255$ )	0.06s	0.39s	0.77s
16x16 image block ( $n = 256, d = 255$ )	0.11s	0.98s	1.98s
32x32 image block ( $n = 1024, d = 255$ )	0.41s	3.91s	7.81s
Exam scores ( $n = 300, d = 100$ )	0.12s	1.14s	2.26s
Salaries ( $n = 30, d = 100000$ )	0.02s	0.13s	0.25s
Ratings ( $n = 5000, d = 4$ )	1.84s	18.5s	37.8s

<b>Application</b>	Lexicographic			Recursive Block		
	Table size	Table time	Enc. time	Table size	Table time	Enc. time
10x10	162 MB	1.81s	0.009s	9 MB	0.87s	0.06s
16x16	1885 MB	13.1s	0.013	32 MB	5.8s	.18s
32x32	fail	-	-	271 MB	316s	3.50s
Exams	988MB	7.17s	0.011s	15 MB	3.81s	.096s
Salaries	4756 MB	79s	0.34s	672 MB	40.5s	4.2s
Ratings	fail	-	-	25 MB	271 s	0.40 s

# Summary

- First proof bounding the mixing time of the Tajik et al. algorithm
- Give practical rank and unrank algorithms for sum-preserving encryption
- Create prototype implementations with performance comparisons



Thank you!