

# Rank Analysis of Cubic Multivariate Cryptosystems

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# Motivation

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# HFE Cryptosystem

- $\mathbb{F}$  a finite prime field of size  $q$ .
- $\mathbb{K}$  field extension of degree  $n$  of  $\mathbb{F}$ .
- $\phi : \mathbb{K} \rightarrow \mathbb{F}^n$  vector space isomorphism.
- $\mathcal{F}(X) = \sum \alpha_{i,j} X^{q^i + q^j} \in \mathbb{K}[X]$
- $S, T$  linear transformations  $\mathbb{F}^n \rightarrow \mathbb{F}^n$ .

## Secret Key

$\mathcal{F}, S$  and  $T$ .

## Public Key

$P = T \circ \phi \circ \mathcal{F} \circ \phi^{-1} \circ S$ , which is given by multivariate **quadratic** polynomials  $f_1, \dots, f_n \in \mathbb{F}[x_1, \dots, x_n]$ .

**Encryption** Evaluation at these polynomials

**Decryption** Inverting  $P$  ( $\mathcal{F}$  is taken as a low degree polynomial)

## Min-Rank Attack (in a nutshell)

1. A symmetric matrix  $(\alpha_{i,j})_{i,j}$  can be associated to  $\mathcal{F}$
  2. This matrix has low rank due to the fact that  $\mathcal{F}$  has low degree
  3. This rank defect is reflected in  $P$  as an instance of the so-called Min-Rank problem
  4. This instance can be solved by practical means
  5. The solution yields valuable information that can be used to recover an equivalent secret key.
- It has been proven that this vulnerability also has a negative impact in the degree of regularity of the system.

The attack seems to require a quadratic setting

- Otherwise no symmetric matrix could be associated to  $\mathcal{F}$

### Countermeasure?

Take the same construction, but with

$$\mathcal{F}(X) = \sum_{0 \leq i \leq j \leq k \leq n-1} \alpha_{i,j,k} X^{q^i+q^j+q^k}.$$

(low degree is still needed for decryption!)

Now the public key is given by **cubic** multivariate polynomials  $f_1, \dots, f_n \in \mathbb{F}[x_1, \dots, x_n]$ .

# Differential attack

Consider the differential  $D_{\mathbf{a}}P(\mathbf{x}) = P(\mathbf{x} + \mathbf{a}) - P(\mathbf{x}) - P(\mathbf{a})$ .

- This differential is composed of quadratic multivariate polynomials. Let  $P'$  be the quadratic homogeneous part.
- We have that  $P' = T \circ \phi \circ \mathcal{F}' \circ \phi^{-1} \circ S$ , where  $\mathcal{F}'$  is the quadratic homogeneous part of  $D_{\mathbf{a}}\mathcal{F}(X)$ .

## The bad news

$\mathcal{F}'$  has the same (low) degree as  $\mathcal{F}$ , so  $P'$  is an instance of quadratic HFE, with the same  $S$  and  $T$ , which is vulnerable to the Min-Rank attack.

# Our Contributions

- We introduce a cubic version of the Min-Rank problem and show how to solve it using natural extensions from the KS modelling.
- We show, experimentally, that taking differentials does not necessarily make the problem easier (as it did in cubic HFE).
- We discuss the implications of a cubic rank defect in the direct algebraic attack.
- We show that cubic big field constructions with a low-rank central polynomial are vulnerable to the cubic Min-Rank attack.

## Related work

- Moody, Perlner, and Smith-Tone do a rank analysis of the cubic ABC scheme.<sup>12</sup>
  - Taking differentials reduces the rank significantly, which allows for a quadratic Min-Rank attack.
  - Their work avoids discussing the rank of cubic polynomials by focusing on the differentials

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<sup>1</sup>Dustin Moody, Ray Perlner, and Daniel Smith-Tone. “Key Recovery Attack on the Cubic ABC Simple Matrix Multivariate Encryption Scheme”. In: *Selected Areas in Cryptography – SAC 2016*. 2017.

<sup>2</sup>Dustin Moody, Ray Perlner, and Daniel Smith-Tone. “Improved Attacks for Characteristic-2 Parameters of the Cubic ABC Simple Matrix Encryption Scheme”. In: *Post-Quantum Cryptography*. 2017.



# Cubic Min-Rank Attack

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## Definition

Let  $A \in \mathbb{F}^{n \times n \times n}$  be a three-dimensional matrix, we define the **rank** of  $A$  as the minimum number of summands  $r$  required to write  $A$  as

$$A = \sum_{i=1}^r \mathbf{u}_i \otimes \mathbf{v}_i \otimes \mathbf{w}_i,$$

where  $\mathbf{u}_i, \mathbf{v}_i, \mathbf{w}_i \in \mathbb{F}^n$ . We denote this number by  $\text{Rank}(A)$ .

- The matrix  $\mathbf{u} \otimes \mathbf{v} \otimes \mathbf{w}$  is defined so that its entry  $(i, j, k)$  is given by  $u_i v_j w_k$ .

- Generalizes the concept of rank for two-dimensional matrices
- It is not trivial to determine the rank of a three-dimensional matrix
  - In fact, the problem is NP-hard, along with many other problems related to three-dimensional rank
- It is not easy to generate three-dimensional matrices with a desired rank
- Determining the maximum rank attainable by a  $n \times n \times n$  matrix remains an open question
  - It is known that this maximum lies between  $\frac{n^2}{3}$  and  $\frac{3n^2}{4}$

### Definition (Cubic Min-Rank Problem)

Given  $M_1, \dots, M_\kappa \in \mathbb{F}^{n \times n \times n}$ , determine whether there exist  $\lambda_1, \dots, \lambda_\kappa \in \mathbb{F}$  such that the rank of  $\sum_{i=1}^{\kappa} \lambda_i M_i$  is less or equal to  $r$ .

- Same definition as in the two-dimensional case but with three-dimensional matrices and using the extended concept of rank.

# Solving the cubic Min-Rank problem

## Theorem (Characterization of rank<sup>3</sup>)

The rank of a matrix  $A \in \mathbb{F}^{n \times n \times n}$  is the minimal number  $r$  of rank one matrices  $S_1, \dots, S_r \in \mathbb{F}^{n \times n}$ , such that, for all slices<sup>4</sup>  $A[i, \cdot, \cdot]$  of  $A$ ,  $A[i, \cdot, \cdot] \in \text{span}(S_1, \dots, S_r)$ .

- Analog in two-dimensional case: the rank is the minimum number of vectors required to span the row space (or the column space).
  - This is the characterization of rank used in the quadratic KS modelling.

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<sup>3</sup>Joseph M Landsberg. *Tensors: geometry and applications*.

<sup>4</sup> $A[i, \cdot, \cdot]$  is the two-dimensional matrix whose entry  $(j, k)$  is given by  $A[i, j, k]$

## Generalization of KS modelling

- Let  $A = \sum_{i=1}^{\kappa} \lambda_i M_i$ .
- Write  $S_i = \mathbf{u}_i \mathbf{v}_i^T$  for some *unknown* vectors  $\mathbf{u}_i, \mathbf{v}_i \in \mathbb{F}^n$ .
- We force the property  $A[i, \cdot, \cdot] \in \text{span}(S_1, \dots, S_r)$ :

$$\sum_{j=1}^r \alpha_{ij} \mathbf{u}_j \mathbf{v}_j^T = A[i, \cdot, \cdot], \text{ for } i = 1, \dots, n.$$

- We get a system of cubic equations
  - # **Variables**  $r(2n) + rn + \kappa$  (entries of the vectors above + linear combination coefficients +  $\lambda_i$ )
  - # **Equations**  $n^3$  ( $n$  equations of  $n \times n$  matrices)

## If $r \ll n$ we can do much better

- It is very likely that  $A[1, \cdot, \cdot], \dots, A[r, \cdot, \cdot]$  are linearly independent, so

$$\text{span}(S_1, \dots, S_r) = \text{span}(A[1, \cdot, \cdot], \dots, A[r, \cdot, \cdot]).$$

- We force the condition  $A[i, \cdot, \cdot] \in \text{span}(A[1, \cdot, \cdot], \dots, A[r, \cdot, \cdot])$  by

$$\sum_{j=1}^r \alpha_{ij} A[j, \cdot, \cdot] = A[i, \cdot, \cdot], \text{ for } i = r + 1, \dots, n.$$

- We get a system of  $n^2(n - r)$  quadratic equations in  $(n - r)r + \kappa$  variables
  - Easier system than the system obtained with the quadratic KS modelling.

# Differentials

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What is the expected rank of the quadratic part of the differential  $D_{\mathbf{a}}f(\mathbf{x}) = f(\mathbf{x} + \mathbf{a}) - f(\mathbf{x}) - f(\mathbf{a})$ , where  $f \in \mathbb{F}[\mathbf{x}]$  is a random homogeneous cubic polynomial of rank  $r$ ?

## Main problem

How to generate random polynomials of a specific rank  $r$ ?

## Definition

We define the symmetric rank of  $S \in \mathbb{F}^{n \times n \times n}$  as the minimum number of summands  $s$  required to write  $S$  as

$$S = \sum_{i=1}^s t_i \mathbf{u}_i \otimes \mathbf{u}_i \otimes \mathbf{u}_i,$$

where  $\mathbf{u}_i \in \mathbb{F}^n$ ,  $t_i \in \mathbb{F}$ . We denote this number by  $\text{SRank}(S)$ .

- It is clear that, in general,  $\text{Rank}(S) \leq \text{SRank}(S)$ .
- $\text{SRank}(S) < \infty$  if  $|\mathbb{F}| \geq 3$ .

### Proposition

Let  $f \in \mathbb{F}[\mathbf{x}]$  be a homogeneous cubic polynomial. If  $g$  is the quadratic homogeneous part of  $Df_{\mathbf{a}}(\mathbf{x})$ , then  $\text{Rank}(g) \leq \text{SRank}(f)$ .

### Proof.

If  $f(\mathbf{x}) = \sum_{i=1}^r t_i u_i(\mathbf{x})u_i(\mathbf{x})u_i(\mathbf{x})$ , then for any  $\mathbf{a} \in \mathbb{F}^n$  the quadratic part of  $Df_{\mathbf{a}}(\mathbf{x})$  is  $\sum_{i=1}^r 3t_i u_i(\mathbf{a})u_i(\mathbf{x})u_i(\mathbf{x})$ . □

## Kruskal Rank

$\text{KRank}(\mathbf{u}_1, \dots, \mathbf{u}_m)$ : maximum integer  $k$  such that any subset of  $\{\mathbf{u}_1, \dots, \mathbf{u}_m\}$  of size  $k$  is linearly independent.

### Theorem (Kruskal Theorem)

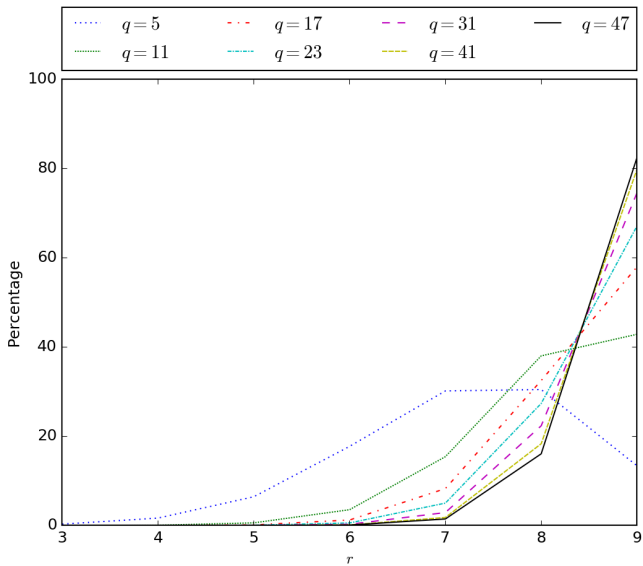
If  $A = \sum_{i=1}^r t_i \mathbf{u}_i \otimes \mathbf{u}_i \otimes \mathbf{u}_i$  and

$$2r + 2 \leq \text{KRank}(t_1 \mathbf{u}_1, \dots, t_r \mathbf{u}_r) + 2 \cdot \text{KRank}(\mathbf{u}_1, \dots, \mathbf{u}_r),$$

then  $\text{Rank}(A) = r$ .

- To generate matrices of rank  $r$ , pick  $\mathbf{u}_1, \dots, \mathbf{u}_r \in \mathbb{F}^n$  and  $t_1, \dots, t_r \in \mathbb{F} - \{0\}$  at random.

$r = 9, n = 20$



# Algebraic Attack

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The complexity of performing a direct algebraic attack (via Groebner bases) is upper bounded by

$$O\left(n^{\omega \frac{r(q-1)+5}{2}}\right),$$

where  $2 \leq \omega \leq 3$  is a linear algebra constant.

- Polynomial in  $n$  if  $r$  and  $q$  are constant.
- Super-polynomial in  $n$  if  $r$  grows with  $n$ .<sup>5</sup>

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<sup>5</sup>This is still an upper bound on the complexity of the attack!

## **Low rank big field constructions**

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- Let  $\mathcal{F} \in \mathbb{K}[X]$  be a homogeneous weight 3 polynomial given by

$$\mathcal{F}(X) = \sum_{1 \leq i, j, k \leq n} \alpha_{i, j, k} X^{q^{i-1} + q^{j-1} + q^{k-1}}$$

- Consider the matrix  $A = (\alpha_{i, j, k})_{i, j, k} \in \mathbb{F}^{n \times n \times n}$ .
- Suppose that  $A$  has low rank  $r$  (e.g. HFE-like construction).
- Let  $A_i$  be the three-dimensional matrix representing the  $i$ -th polynomial of the public key  $T \circ \phi \circ \mathcal{F} \circ \phi^{-1} \circ S$ .

- Consider the trilinear form  $\mathcal{T} : \mathbb{K}^n \times \mathbb{K}^n \times \mathbb{K}^n \rightarrow \mathbb{K}$  given by

$$\mathcal{T}(\beta, \delta, \gamma) = \sum_{1 \leq i, j, k \leq n} \alpha_{i, j, k} \cdot (\beta_i \delta_j \gamma_k).$$

## Theorem

*There exist  $\lambda_i \in \mathbb{K}$  such that  $\sum_{i=1}^n \lambda_i A_i = A'$ , where  $A'$  is the three-dimensional matrix representing the trilinear form  $\mathcal{T} \circ (\Delta S)$ .<sup>6</sup>*

- We can prove that  $\text{Rank}(A') \leq \text{Rank}(A)$
- We obtain an instance of the cubic Min-Rank problem
- Equivalent secret keys

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<sup>6</sup> $\Delta \in \mathbb{K}^{n \times n}$  is a matrix associated to the field extension  $\mathbb{K}$  over  $\mathbb{F}$

# Conclusions

- Rank weaknesses are still present in the cubic setting
- Instances of the cubic Min-Rank problem can be solved
  - More efficiently than in the quadratic setting for  $r \ll n$ .
  - Solving a cubic system for  $r \geq n$ .
- Taking differentials does not allow, in general, to transform the problem into a quadratic one that is easier.
- Low, fixed rank constructions cannot be secure
  - The system is distinguishable from random
  - Susceptible to Min-Rank attack (obtaining equivalent secret keys)
  - Makes direct algebraic attack polynomial

## Future Work

- Finding other algorithms to solve the cubic Min-Rank problem (e.g. generalization of minors modelling)
- Solving the Min-Rank problem in the setting of characteristic 2 and 3
- Developing new encryption/signature schemes with low enough rank to allow decryption/signing but large enough rank to avoid the Min-Rank attack
- Using the hardness of three-dimensional rank problems as a security assumption

Thanks