### Latest Developments in Rainbow Cryptanalysis



Giovanni Tognolini

University of Trento

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# Main objects of multivariate cryptography

$$p(x_1,...,x_n) = \sum_{i=1}^n \sum_{j=1}^n p_{ij} \cdot x_i x_j + \sum_{i=1}^n p_i \cdot x_i + p_0$$

# Why are these polynomials so important?

Given m multivariate quadratic polynomials

$$\mathcal{P}(x) := \begin{pmatrix} p^{(1)}(x) \\ \vdots \\ p^{(m)}(x) \end{pmatrix}$$

it is difficult to solve the system  $\mathcal{P}(x) = y$ .

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It is a good starting point to construct cryptosystems!

## How do we construct cryptosystems?

We take an easily invertible quadratic map  $\mathcal{F}: \mathbb{F}_q^n \longrightarrow \mathbb{F}_q^m$ To hide its structure, we hide it with two invertible affine maps  $\mathcal{S}: \mathbb{F}_q^m \longrightarrow \mathbb{F}_q^m$  and  $\mathcal{T}: \mathbb{F}_q^n \longrightarrow \mathbb{F}_q^n$ .

$$\begin{split} \mathrm{PK} : (\mathcal{P} := \mathcal{S} \circ \mathcal{F} \circ \mathcal{T}) \\ \mathrm{SK} : (\mathcal{S}, \mathcal{F}, \mathcal{T}) \end{split}$$

$$h(m) \in \mathbb{F}_q^m \xrightarrow{\quad \mathcal{S}^{-1} \quad} x \in \mathbb{F}_q^m \xrightarrow{\quad \mathcal{F}^{-1} \quad} y \in \mathbb{F}_q^n \xrightarrow{\quad \mathcal{T}^{-1} \quad} z \in \mathbb{F}_q^n$$

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Two famous multivariate cryptosystems: (U)OV and Rainbow

# (U)OV traditional description

#### Idea

We contruct an easily invertible  $\mathcal{F}$  and hide its structure.

- Consider a finite field  $\mathbb{F}_q$ .
- Let  $v, o \in \mathbb{Z}$  and define n := v + o.
- Let  $V := \{1, ..., v\}$  and  $O := \{v+1, ..., n\}$ . We will call  $x_1, ..., x_v$  the vinegar variables, and  $x_{v+1}, ..., x_n$  the oil variable.
- The central map  $\mathcal{F}: \mathbb{F}_q^n \longrightarrow \mathbb{F}_q^o$  consists of o quadratic polynomials of the form

$$f^{(k)} = \sum_{i,j \in V} \alpha_{i,j}^{(k)} x_i x_j + \sum_{i \in V, j \in O} \beta_{i,j}^{(k)} x_i x_j$$

#### Observation

 $f^{(k)}$  contains no quadratic terms  $x_i x_j$  with both  $i, j \in O$ .

## How to invert $\mathcal{F}$

### Example

- $\mathbb{F} = \mathbb{F}_7$ .
- $\bullet$   $o = v = 2 (\longrightarrow OV)$
- Let  $\mathcal{F}$  be given by

$$f^{(1)}(x_1, ..., x_4) = 2x_1^2 + 3x_1x_2 + 6x_1x_3 + x_1x_4 + 4x_2^2 + 5x_2x_4$$
  
$$f^{(2)}(x_1, ..., x_4) = 3x_1^2 + 6x_1x_2 + 5x_1x_4 + 3x_2^2 + 5x_2x_3 + x_2x_4$$
  
$$\vdots$$

Suppose we want to find a preimage of (3, 4).



We proceed as follows

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- 1. Choose random value for the vinegar variables, e.g.  $(x_1, x_2) = (1, 4)$ .
- 2. Substitute them into  $f^{(1)}$  and  $f^{(2)}$ :

$$f_{|_{(1,4)}}^{(1)} = 6x_3 + 1$$

$$f_{|_{(1,4)}}^{(2)} = 5 + 2x_4 - x_3$$

3. Solve the *linear* system. We obtain  $(x_3, x_4) = (5, 2)$ 

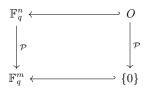


The required preimage is x = (1, 4, 5, 2).

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# (U)OV alternative description

- The public key is a multivariate quadratic map  $\mathcal{P}: \mathbb{F}_q^n \longrightarrow \mathbb{F}_q^m$  which vanishes on a secret linear subspace O of dimension m.
- The private key is a description of O.



### How to generate a public key?

- Pick the subspace O uniformly at random.
- $\bullet$  Pick  $\mathcal{P}$  uniformly at random.

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## Before finding preimages...

### Observation (Polar form of $\mathcal{F}$ )

Given a multivariate quadratic polynomial p(x) we can define its polar form

$$p'(x, y) := p(x + y) - p(x) - p(y) + p(0)$$

Similarly, given m multivariate quadratic polynomials, we define

$$\mathcal{P}'(x,y) := \begin{pmatrix} p'_1(x,y) \\ \vdots \\ p'_m(x,y) \end{pmatrix}$$

 $\mathcal{P}'(x,y)$  is a simmetric and bilinear map.

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## Finding preimages

Suppose we want to find a preimage for  $t \in \mathbb{F}_q^m$ .

- Pick  $v \in \mathbb{F}_q^n$  randomly.
- Solve  $\mathcal{P}(v+o) = t$  for  $o \in O$ .

### Observation (This is an easy task!)

Memo: 
$$\mathcal{P}'(x,y) := \mathcal{P}(x+y) - \mathcal{P}(x) - \mathcal{P}(y)$$

$$\mathcal{P}(v+o) = \mathcal{P}(v) + \mathcal{P}(o) + \mathcal{P}'(v,o) = t$$

- $\mathcal{P}(v)$  is fixed.
- $\mathcal{P}(o) = 0$ .
- $\mathcal{P}'(v, o)$  is linear and it has
  - ightharpoonup m variables.
  - $\triangleright$  m equations.

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## Rainbow

Is just a multi-layer version of UOV.

### Advantages

- Smaller key size.
- Smaller signature size.
- Better performance.

# Approaching Rainbow with an example

### Example

- $\mathbb{F} = \mathbb{F}_7$ .
- Let  $\mathcal{F}$  be given by

$$f^{(1)} = x_1^2 + 3x_1x_2 + 5x_1x_3 + 6x_1x_4 + 2x_2^2 + 6x_2x_3 + 4x_2x_4$$

$$f^{(2)} = 2x_1^2 + x_1x_2 + x_1x_3 + 3x_1x_4 + x_2^2 + x_2x_3 + 4x_2x_4$$

$$f^{(3)} = 2x_1^2 + 3x_1x_2 + 3x_1x_3 + 3x_1x_4 + x_1x_5 + 3x_1x_6 + 4x_2^2 + x_2x_3 + 4x_2x_4 + x_2x_5 + 3x_2x_6 + 3x_3x_4 + x_3x_5 + 2x_3x_6 + 3x_4x_5$$

$$f^{(4)} = 2x_1^2 + 5x_1x_2 + x_1x_3 + 5x_1x_4 + 5x_1x_6 + 5x_2^2 + 3x_2x_3 + 5x_2x_5 + 4x_2x_6 + 3x_3^2 + 5x_3x_4 + 4x_3x_5 + 2x_3x_6 + x_4^2 + 6x_4x_5 + 3x_4x_6$$

$$\downarrow$$

Suppose we want to find a preimage of (6, 2, 0, 5).

### We proceed as follows

- 1. Choose random values for  $x_1$  and  $x_2$ , e.g.  $(x_1, x_2) = (0, 1)$ .
- 2. Substitute them into  $f^{(1)}, ..., f^{(4)}$ :

$$f_{\begin{vmatrix} (0,1) \\ (0,1) \end{vmatrix}}^{(1)} = 6x_3 + 5x_4 + 2$$

$$f_{\begin{vmatrix} (0,1) \\ (0,1) \end{vmatrix}}^{(2)} = x_3 + 4x_4 + 1$$

$$f_{\begin{vmatrix} (0,1) \\ (0,1) \end{vmatrix}}^{(3)} = 3x_4x_5 + 2x_3x_6 + x_3x_5 + 3x_3x_4 + 3x_6 + x_5 + 4x_4 + x_3 + 4$$

$$f_{\begin{vmatrix} (0,1) \\ (0,1) \end{vmatrix}}^{(4)} = 3x_4x_6 + 6x_4x_5 + x_4^2 + 2x_3x_6 + 4x_3x_5 + 5x_3x_4 + 3x_3^2 + 4x_6 + 5x_5 + 3x_3 + 5$$

3. Solve the small *linear* system and obtain  $(x_3, x_4) = (5, 6)$ .

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4. Substitute these values into  $f^{(3)}, f^{(4)}$ :

$$f_{|_{(0,1,5,6)}}^{(3)} = -4x_5 - 7x_6 - 2$$

$$f_{|_{(0,1,5,6)}}^{(4)} = -9x_5 - 3x_6 + 1$$

5. Solve the *linear* system and obtain  $(x_5, x_6) = (3, 6)$ 



The required preimage is x = (0, 1, 5, 6, 3, 6).

## Rainbow traditional description

We don't really care about it here...

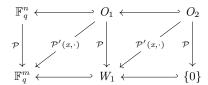
## Rainbow alternative description

- The public key is a multivariate quadratic map  $\mathcal{P}: \mathbb{F}_q^n \longrightarrow \mathbb{F}_q^m$ .
- The private key consists of
  - ► Two sequences of nested subspaces:

$$\mathbb{F}_q^n = O_0 \supseteq O_1 \supseteq O_2$$
  
$$\mathbb{F}_q^m = W_0 \supseteq W_1 \supseteq W_2 = \{0\}$$

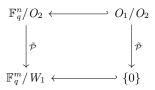
- ► Some constrains:
  - $\star$  dim $(O_i)$  = dim $(W_{i-1})$ .

  - \*  $\mathcal{P}(x) \in W_i$  for all  $x \in O_i$ . \*  $\mathcal{P}'(x, y) \in W_{i-1}$  for all  $x \in \mathbb{F}_q^n, y \in O_i$ .



## How to sign?

- Suppose we have a target  $t \in \mathbb{F}_q^m$ .
- Consider the UOV instance



### Observation

This is indeed an UOV instance as  $\dim(O_1/O_2) = \dim(\mathbb{F}_q^m/W_1)$ 

Pick  $[v] \in \mathbb{F}_q^n/O_2$  randomly and solve for  $[o_1] \in O_1/O_2$  the system

$$\tilde{\mathcal{P}}([v] + [o_1]) = [t]$$

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• Solve for  $o_2 \in O_2$  the system

$$\mathcal{P}(v + o_1 + o_2) = t$$

### Observation (This is an easy task!)

Memo: 
$$\mathcal{P}'(x, y) := \mathcal{P}(x + y) - \mathcal{P}(x) - \mathcal{P}(y)$$

$$\mathcal{P}((v+o_1)+o_2) = \mathcal{P}(v+o_1) + \mathcal{P}(o_2) + \mathcal{P}'(v+o_1,o_2) = t$$

- $\mathcal{P}(v+o_1)$  is fixed.
- $\mathcal{P}(o_2) = 0$ .
- $\mathcal{P}'(v+o_1,o_2)$  is linear and it has
  - $ightharpoonup \dim(O_2) = \dim(W_1)$  variables.
  - $ightharpoonup \dim(W_1)$  equations.

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Attacking Rainbow

# Simple attack

### Idea

0. Consider the map

$$D_x \colon \mathbb{F}_q^n \longrightarrow \mathbb{F}_q^m \\ y \longmapsto \mathcal{P}'(x,y) .$$

- 1. Observe that with high probability it has an element in  $O_2 \cap \ker(D_x)$ .
- 2. Find this element.
- 3. Reconstruct  $O_2$ .
- 4. Reconstruct W.
- 5. Reconstruct  $O_1$  (key recovery) or forge a signature.

1.  $D_x$  has an element in  $O_2 \cap \ker(D_x)$  with high probability.

### Observation

•  $D_x$  is a linear map, indeed

$$D_x(y_1 + y_2) = \mathcal{P}'(x, y_1 + y_2)$$
  
=  $\mathcal{P}'(x, y_1) + \mathcal{P}'(x, y_2)$   
=  $D_x(y_1) + D_x(y_2)$ .

•  $D_{x|_{O_2}}: O_2 \longrightarrow W$ , indeed

$$D_x(o) = \mathcal{P}'(x, o) \in W.$$

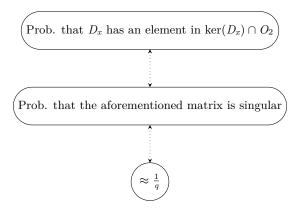
 $\bullet \dim(O_2) = o_2 = \dim(W).$ 

#### Result

We can represent  $D_{x|_{O_2}}$  as a square  $o_2$ -by- $o_2$  random matrix over  $\mathbb{F}_q$ .

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## 1. $D_x$ has an element in $O_2 \cap \ker(D_x)$ with high probability.



2. Find this element  $o \in \ker(D_x) \cap O_2$ .

A good idea would be to solve the system

$$\begin{cases} D_x(o) = 0 \\ \mathcal{P}(o) = 0 \end{cases}$$

#### Observation

- $D_x(o) = 0$  consists of m linear equations in the n variables of o.
- $\mathcal{P}(o) = 0$  consists of m homogeneous quadratic equations in the n variables of o.

We can reduce to a system of

- $\bullet$  *m* homogeneous equations.
- n-m variables.

2. Find this element  $o \in \ker(D_x) \cap O_2$ .

### Concreterly:

Let  $B \in \mathbb{F}_q^{n \times (n-m)}$  a basis for  $\ker(D_x)$ .

$$\begin{cases} D_x(o) = 0 \\ \mathcal{P}(o) = 0 \end{cases} \iff \begin{cases} o \in \ker(D_x) \\ \mathcal{P}(o) = 0 \end{cases} \iff \begin{cases} o = By \\ \mathcal{P}(o) = 0 \end{cases} \iff \tilde{\mathcal{P}}(y) := \mathcal{P}(By) = 0$$

### Observation

Finding  $o \in \ker(D_x) \cap O_2$  reduces to find  $y \in \mathbb{F}_q^{n-m}$  s.t.  $\tilde{\mathcal{P}}(y) = 0$ .

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2. Find this element  $o \in \ker(D_x) \cap O_2$ .

We would like to solve  $\tilde{\mathcal{P}}(y) = 0$  with the XL algorithm.

### Observation

In order to apply the XL algorithm we need to be sure that the system is random.

We distinguish the cases:

- $\operatorname{ch}(\mathbb{F}_q)$  odd.
- $\operatorname{ch}(\mathbb{F}_q)$  even.

### Odd characteristic

### Observation

In this case  $\tilde{\mathcal{P}}$  behaves like a random system.

What does it mean that it behaves like a random system?

The ranks of Macaulay matrices (at various degree D) of  $\tilde{\mathcal{P}}(x) = 0$  are identical to the ranks of systems of uniformly random quadratic equations with the same dimensions.

### Conclusion

If a solution to  $\tilde{\mathcal{P}}(x) = 0$  exists, we can find it with XL.

### Even characteristic

### Observation

In this case  $\tilde{\mathcal{P}}$  does not behave like a random system.



Applying the XL sometimes fails.

### Why?

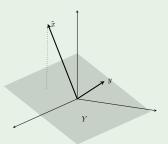
In characteristic 2 it is possible to show that there is an  $\tilde{x}$  (known to the attacker) such that:

$$\tilde{\mathcal{P}}(\tilde{x}+y) = \tilde{\mathcal{P}}(\tilde{x}) + \tilde{\mathcal{P}}(y)$$

which is not something that usually happens for random  $\tilde{\mathcal{P}}$ .

## How to solve this problem?

- Restrict  $\tilde{\mathcal{P}}$  to  $Y \subseteq \mathbb{F}_q^{n-m}$  (Y is a subspace of dimension n-m-1 that does not contain  $\tilde{x}$ ).
- Find  $y \in Y$  s.t.  $\tilde{\mathcal{P}}(y) = \alpha \tilde{\mathcal{P}}(\tilde{x})$ .



Why are we looking for such an y?

In this case  $\tilde{x} + \alpha^{-\frac{1}{2}}y$  is a solution to  $\tilde{\mathcal{P}}(x) = 0$ , indeed:

$$\begin{split} \tilde{\mathcal{P}}(\tilde{x} + \alpha^{-\frac{1}{2}}y) &= \tilde{\mathcal{P}}(\tilde{x}) + \tilde{\mathcal{P}}(\alpha^{-\frac{1}{2}}y) \\ &= \tilde{\mathcal{P}}(\tilde{x}) + \alpha^{-1}\tilde{\mathcal{P}}(y) \\ &= 0 \end{split}$$

How do we find y?

$$\tilde{\mathcal{P}}(y) = \alpha \tilde{\mathcal{P}}(\tilde{x}) \iff \left\{ \tilde{p}_i(y) = \alpha \tilde{p}_i(\tilde{x}) \right\}_{i=1}^m \\ \iff (\star)$$

#### Observation

If we assume (with loss of generality) that  $\tilde{p}_1(\tilde{x}) \neq 0$  then we can write

$$\alpha = \frac{\tilde{p}_1(y)}{\tilde{p}_1(\tilde{x})}$$

and we obtain

$$(\star) \iff \left\{ \tilde{p}_i(y) = \frac{\tilde{p}_1(y)}{\tilde{p}_1(\tilde{x})} \cdot \tilde{p}_i(\tilde{x}) \right\}_{i=2}^m \\ \iff \left\{ \tilde{p}_i(y) \tilde{p}_1(\tilde{x}) - \tilde{p}_1(y) \tilde{p}_i(\tilde{x}) = 0 \right\}_{i=2}^m$$

So...

In order to find y we restrict  $\tilde{\mathcal{P}}$  to Y and solve the previous system.

### Result

The new system

$$\left\{\tilde{p}_i(y)\tilde{p}_1(\tilde{x}) - \tilde{p}_1(y)\tilde{p}_i(\tilde{x}) = 0\right\}_{i=2}^m$$

is a system of

- m-1 homogeneous quadratic equations.
- n-m-1 variables.

and it behaves like a random system.

### Conclusion

If a solution exists, we can find it with XL.



At this point we managed to find an element  $o \in O_2$ .

 $\downarrow$ 

It is now easy to recover  $O_2$  and W.

## Recovering W

#### Observation

Given a single vector  $o \in O_2$ , we can compute

$$\langle \mathcal{P}'(o, e_1), ..., \mathcal{P}'(o, e_n) \rangle \subseteq W$$

which will be (with overwhelming probability) an equality.

### Recovering $O_2$

• Let V a change of variables which sends W to the last  $o_2$  coordinates of  $\mathbb{F}_q^m$ .

$$V: \quad \mathbb{F}_q^m \xrightarrow{} \quad \mathbb{F}_q^m$$

$$w \in W \longmapsto (0, 0, ..., 0, \star, ..., \star)^T$$

• We can split up  $V \circ \mathcal{P}$  as

$$V \circ \mathcal{P}(x) = \begin{cases} \mathcal{P}_1(x) & \longleftarrow \text{ first } m - o_2 \text{ coordinates} \\ \mathcal{P}_2(x) & \longleftarrow \text{ last } o_2 \text{ coordinates} \end{cases}$$

### Observation

With very high probability

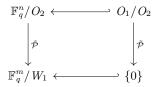
$$O_2 = \ker \left( x \mapsto \begin{pmatrix} \mathcal{P}'_1(e_1, x) \\ \vdots \\ \mathcal{P}'_1(e_n, x) \end{pmatrix} \right)$$

- " $\subseteq$ " is clear:  $o \in O_2 \Longrightarrow \mathcal{P}'_1(e_i, o) = V \circ \mathcal{P}'(e_i, o)| = 0.$
- "=" is very likely.

We can reduce Rainbow to a UOV instance with parameters  $n' = n - o_2$   $m' = m - o_2$ .

# An old slide: How to sign?

- We had a target  $t \in \mathbb{F}_q^m$ .
- We considered the UOV instance



We picked  $[v] \in \mathbb{F}_q^n/O_2$  randomly and solved for  $[o_1] \in O_1/O_2$  the system

$$\tilde{\mathcal{P}}([v] + [o_1]) = [t]$$

• We solved for  $o_2 \in O_2$  the system

$$\mathcal{P}(v + o_1 + o_2) = t$$

These are easy tasks!

# Concluding the Simple attack

Two ways to force the first step:

- 1. Recover  $O_1$  (full key recovery).
- 2. Forge the signature.

#### Observation

- SL1 parameter sets of 2° and 3° round NIST submission:
  - (n', m') = (64, 32)(n', m') = (68, 32)

  - $\rightarrow$  Key recovery with Kipnis-Shamir attack  $(q^{n'-2m'} \cdot \text{poly}(n))$
- SL3 and 5
  - (n', m') is too big for a full key recovery
  - $\rightarrow$  We can find a preimage for a UOV instance with XL.

# Performance on NIST parameters sets

Parameter set		$(q, n, m, o_2)$	Simple attack	Known attacks
(2°)	SL1	(16,96,64,32)	61*	123
	SL3	(256, 140, 72, 36)	186	151
	SL5	(256, 188, 96, 48)	246	191
(3°)	SL1	(16, 100, 64, 32)	69*	127
	SL3	(256, 148, 80, 48)	160	177
	SL5	(256, 196, 100, 64)	257	226

We can try to combine the simple attack with some known attack

# Rectangular MinRank Attack

### Rectangular MinRank Problem

An instance of this problem is:

- a list of matrices  $L_1, ..., L_k$ .
- a target rank r.

The task is to find a non-zero linear combination of the matrices whose rank is at most r.

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#### Consider the n matrices

$$L_i := egin{pmatrix} \mathcal{P}'(e_1, e_i) \ dots \ \mathcal{P}'(e_n, e_i) \end{pmatrix}$$

#### Observation

• Since  $\mathcal{P}$  is bilinear

$$\forall x \in \mathbb{F}_q^n$$
 
$$\sum_{i=1}^n x_i L_i = \begin{pmatrix} \mathcal{P}'(e_1, x) \\ \vdots \\ \mathcal{P}'(e_n, x) \end{pmatrix}$$

Furthermore

$$o \in O_2 \Longrightarrow \mathcal{P}'(e_i, o) \in W$$
  
 $\Longrightarrow \operatorname{rank}(\sum o_i L_i) \le \dim W$ 

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So...

We have n matrices  $\{L_i\}_{i=1}^n$ We know there exists an x s.t.  $\operatorname{rank}(\sum x_i L_i) \leq \dim W$ 

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This is an instance of the MinRank problem!

,

We can use known algorithm to solve this problem (if o is a solution, then with overwhelming prob.  $o \in O_2$ )

#### Observation

As before, once a solution  $o \in O_2$  is found, the security of Rainbow is reduced to the security of a UOV instance with

$$\begin{cases} n' = n - o_2 \\ m' = m - o_2 \end{cases}$$

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Combination of previous attacks

The idea is to solve (for  $o \in \mathbb{F}_q^n$ )

min  
Rank on 
$$\sum o_i L_i$$

but since we expect to have  $o \in O_2$ , we add the constrain

$$o \in \ker(D_x)$$

for a random  $x \in \mathbb{F}_q^n$ .

### Why?

 $\ker(D_x) \cap O_2 \neq \{0\}$  with prob.  $\approx \frac{1}{q}$ .

#### CONS

We have to repeat the attack on average approx. q times.

#### **PROS**

Now we have a minRank problem with only n-m matrices (definitely less than the original minRank problem).

# Performance on NIST parameters sets

Parameter set		$(q,n,m,o_2)$	Combined attack	Known attacks
(2°)	SL1	(16,96,64,32)	93*	123
	SL3	(256, 140, 72, 36)	131	151
	SL5	(256, 188, 96, 48)	164	191
(3°)	SL1	(16, 100, 64, 32)	99*	127
	SL3	(256, 148, 80, 48)	157	177
	SL5	(256, 196, 100, 64)	206	226

### Conclusions

We could move to larger parameters, BUT:

- The new signature and public keys would be very big.
- These seems to be room for improvement for attacks.
- 3 The resulting Rainbow scheme would be less efficient than UOV.



There is no reason to prefer Rainbow over UOV, since:

- Rainbow is based on UOV.
- UOV is older.
- UOV is simpler.
- UOV has a smaller attack surface.



### Bibliography



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#### XL

We would like to solve  $\tilde{\mathcal{P}}(y) = 0$  with the XL algorithm.

### Memo: XL algorithm

It solves an instance with m random homogeneous equations in n variables at the cost of

$$3\binom{n-1+D}{D}^2\binom{n+1}{2}$$

field multiplication, where D is the operating degree of XL.

#### Observation

D is the smallest integer such that the coefficient of the  $t^D$  term in the power expansion of

$$\frac{(1-t^2)^m}{(1-t)^n}$$

is non positive.

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#### Example

Suppose we want to find a solution to a system of 63 random homogeneous quadratic equations in 31 variables. We have:

$$\frac{(1-t^2)^{63}}{(1-t)^{31}} = 1 + 31t + 433t^2 + 3503t^3 + 17081t^4 + 41447t^5 - 44919t^6 + O(t^7)$$

so we can run XL at degree D=6, with an estimated cost of

$$3\binom{31-1+6}{6}^2\binom{31+1}{2} \approx 2^{52.3}$$

field multiplications.

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