

Multivariate Quadratic Public-Key Cryptography Part 3: Small Field Schemes

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Oil-Vinegar Polynomials [Patarin 1997]

Let \mathbb{F} be a (finite) field. For $o, v \in \mathbb{N}$ set $n = o + v$ and define

$$p(x_1, \dots, x_n) = \underbrace{\sum_{i=1}^v \sum_{j=i}^v \alpha_{ij} \cdot x_i \cdot x_j}_{v \times v \text{ terms}} + \underbrace{\sum_{i=1}^v \sum_{j=v+1}^n \beta_{ij} \cdot x_i \cdot x_j}_{v \times o \text{ terms}} + \underbrace{\sum_{i=1}^n \gamma_i \cdot x_i}_{\text{linear terms}} + \delta$$

x_1, \dots, x_v : Vinegar variables x_{v+1}, \dots, x_n : Oil variables, no $o \times o$ terms.

If we randomly set x_1, \dots, x_v , result is linear in x_{v+1}, \dots, x_n

$v \times v$ terms $v \times o$ terms $o \times o$ terms v terms o terms

quadratic	quadratic	0	linear in v	linear in o	δ
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Oil-Vinegar Polynomials (2)

Let $\tilde{p}(x_1, \dots, x_n)$ be

(Unbalanced) Oil-Vinegar matrix

\tilde{p} the homogeneous quadratic part of $p(x_1, \dots, x_n)$ can be written as quadratic form $\tilde{p}(\mathbf{x}) = \mathbf{x}^T \cdot M \cdot \mathbf{x}$ with

$$M = \left(\begin{array}{c|c} *_{v \times v} & *_{o \times v} \\ \hline *_{v \times o} & 0_{o \times o} \end{array} \right)$$

where $*$ denotes arbitrary entries subject to symmetry.

Inversion of the UOV central map

Each central polynomial has the form

$v \times v$ terms $v \times o$ terms $o \times o$ terms v terms o terms

quadratic	quadratic	0	linear in v	linear in o	δ
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Inversion of the central map

Each central polynomial has the form

$v \times v$ terms	$v \times o$ terms	$o \times o$ terms	v terms	o terms	
quadratic	quadratic	0	linear in v	linear in o	δ

Choose random values for the Vinegar variables x_1, \dots, x_v

$v \times v$ terms	$v \times o$ terms	$o \times o$ terms	v terms	o terms	
constant	linear in o	0	constant	linear in o	δ

\Rightarrow Linear equation in the o Oil variables

Inversion of the central map (2)

Let each of o components of a UOV central map be a UOV polynomial.

After guessing Vinegar variables

When we guess the Vinegar variables x_1, \dots, x_v , we get o linear equations in the o Oil variables $x_{v+1}, \dots, x_n \Rightarrow$ recovered by (Gaussian) elimination

If the system has no solution?

Just choose other values for the Vinegar variables x_1, \dots, x_v and try again.

Inversion of the central map (2)

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Toy Example in $\mathbb{F} = \text{GF}(7)$ with $o = v = 2$

- $\mathcal{Q} = (f^{(1)}, f^{(2)})$ with

$$f^{(1)}(\mathbf{x}) = 2x_1^2 + 3x_1x_2 + 6x_1x_3 + x_1x_4 + 4x_2^2 + 5x_2x_4 + 3x_1 + 2x_2 + 5x_3 + x_4 + 6,$$
$$f^{(2)}(\mathbf{x}) = 3x_1^2 + 6x_1x_2 + 5x_1x_4 + 3x_2^2 + 5x_2x_3 + x_2x_4 + 2x_1 + 5x_2 + 4x_3 + 2x_4 + 1.$$

- Goal: Find a pre image $\mathcal{Q}^{-1}(\mathbf{y})$, $\mathbf{y} = (3, 4)$
- Choose random values for x_1 and x_2 , e.g. $(x_1, x_2) = (1, 4)$
- $\tilde{f}^{(1)}(x_3, x_4) = 4x_3 + x_4 + 4 = w_1 = 3$, $\tilde{f}^{(2)}(x_3, x_4) = 3x_3 + 4x_4 = w_2 = 4$
- The pre image of \mathbf{y} is $\mathbf{x} = (1, 4, 1, 2)$.

Operations of UOV

Key Generation

Take a UOV central map \mathcal{Q} and invertible $\mathcal{S} : \mathbb{F}^n \rightarrow \mathbb{F}^n$. $\mathcal{P} = \mathcal{Q} \circ \mathcal{S}$.

Signature Generation

- 1 Given: message d , take its hash $\mathbf{y} = \mathcal{H}(d)$ under $\mathcal{H} : \{0, 1\}^* \rightarrow \mathbb{F}^o$.
- 2 Compute a pre-image $\mathbf{x} \in \mathbb{F}^n$ of \mathbf{y} under the central map \mathcal{Q}
 - ▶ Choose random values for the Vinegar variables x_1, \dots, x_v and substitute them into the central map polynomials $f^{(1)}, \dots, f^{(o)}$
 - ▶ Solve the resulting linear system for the Oil variables x_{v+1}, \dots, x_n
 - ▶ If the system has no solution, choose other values for the Vinegar variables and try again.
- 3 Compute the signature $\mathbf{w} \in \mathbb{F}^n$ by $\mathbf{w} = \mathcal{S}^{-1}(\mathbf{x})$.

Operations of UOV

Key Generation

Take a UOV central map Q and invertible $S : \mathbb{F}^n \rightarrow \mathbb{F}^n$. $\mathcal{P} = Q \circ S$.

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- 2 Compute a pre-image $\mathbf{x} \in \mathbb{F}^n$ of \mathbf{y} under the central map Q
- 3 Compute the signature $\mathbf{w} \in \mathbb{F}^n$ by $\mathbf{w} = S^{-1}(\mathbf{x})$.

Signature Verification

Given: message d , signature $\mathbf{w} \in \mathbb{F}^n$

- 1 Compute $\mathbf{z} = \mathcal{H}(d)$.
- 2 Compute $\mathbf{z}' = \mathcal{P}(\mathbf{w})$.

Accept the signature $\Leftrightarrow \mathbf{z} = \mathbf{z}'$

Kipnis-Shamir OV attack when $o = v$

$$\mathcal{O} := \{\mathbf{x} \in \mathbb{F}^n : x_1 = \dots = x_v = 0\} \quad \text{“Oilspace”}$$

$$\mathcal{V} := \{\mathbf{x} \in \mathbb{F}^n : x_{v+1} = \dots = x_n = 0\} \quad \text{“Vinegarspace”}$$

Let E, F be invertible “OV-matrices”, i.e. $E, F = \begin{pmatrix} \star & \star \\ \star & 0 \end{pmatrix}$ Then

$E \cdot \mathcal{O} \subset \mathcal{V}$. Since the two has the same rank, equality holds, so $(F^{-1} \cdot E) \cdot \mathcal{O} = \mathcal{O}$, i.e. \mathcal{O} is an invariant subspace of $F^{-1} \cdot E$.

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Common Subspaces

Let H_i be the matrix representing the homogeneous quadratic part of the i -th public polynomial. Then we have $H_i = S^T \cdot E_i \cdot S$, i.e. $S^{-1}(\mathcal{O})$ is an invariant subspace of the matrix $(H_j^{-1} \cdot H_i)$, and we find S^{-1} .

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Summary of the Standard UOV Attack

- for $v \leq o$, breaks the balanced OV scheme in polynomial time.
- For $v > o$ the complexity of the attack is about $q^{v-o} \cdot o^4$.

\Rightarrow Choose $v \approx 2 \cdot o$ (unbalanced Oil and Vinegar (UOV)) [KP99]

What happens when $v > o$?

Invariant Subspaces

- $E \cdot \mathcal{O} \subset \mathcal{V}$, where E is a UOV matrix.
- If E^{-1} exists then $E^{-1}\mathcal{V}$ is a v -dimensional subspace containing \mathcal{O} .
- If E, F are two invertible UOV matrices, then $FE^{-1}\mathcal{O}$ is the same dimension as \mathcal{O} and both lies in $F^{-1}\mathcal{V}$. So $I = \mathcal{O} \cap F^{-1}E\mathcal{O}$ is dimension at least $2o - v$.
- I is mapped by $F^{-1}E$ into a subspace $F^{-1}E\mathcal{O}$, of dimension o . The probability for a non-zero vector to be mapped to its own multiple is $(q-1)/(q^d-1)$. The expected value is the number of non-zero vectors times this probability divided by $q-1$ (since each eigenvector is counted $q-1$ times), or $(q^{2o-v}-1)/(q^o-1) \sim q^{-(v-o)}$.
- So 1 out of $\sim q^{(v-o)}$, \mathcal{O} contains an invariant subspace of $F^{-1}E$.

Choose an H_i which is invertible and take an arbitrary linear combination $H = \sum_i \alpha_i H_i$, $H_i^{-1}H$ with probability $q^{-(v-o)}$ contains an invariant subspace which is in $S^{-1}\mathcal{O}$. Repeat o times to obtain the entire $S^{-1}\mathcal{O}$.

Other Attacks

- **Collision Attack:** $o \geq \frac{2^{2\ell}}{\log_2(q)}$ for ℓ -bit security.
- **Direct Attack:** Try to solve the public equation $\mathcal{P}(\mathbf{w}) = \mathbf{z}$ as an instance of the MQ-Problem. The public systems of UOV behave much like random systems, but they are highly underdetermined ($n = 3 \cdot m$)

Result [Thomae]: A multivariate system of m equations in $n = \omega \cdot m$ variables can be solved in the same time as a determined system of $m - \lfloor \omega \rfloor + 1$ equations.

$\Rightarrow m$ has to be increased by 2.

Other Attacks

- **Collision Attack:** $o \geq \frac{2^{2\ell}}{\log_2(q)}$ for ℓ -bit security.
- **Direct Attack:** Try to solve the public equation $\mathcal{P}(\mathbf{w}) = \mathbf{z}$ as an instance of the MQ-Problem. The public systems of UOV behave much like random systems, but they are highly underdetermined ($n = 3 \cdot m$) \Rightarrow m has to be increased by 2.
- **UOV-Reconciliation attack:** Try to find a linear transformation S (“good keys”) which transforms the public matrices H_i into the form of UOV matrices

$$(S^T)^{-1} \cdot H_i \cdot S^{-1} = \begin{pmatrix} \star & \star \\ \star & 0 \end{pmatrix}, \quad S = \begin{pmatrix} 1 & \star \\ 0 & 1 \end{pmatrix}$$

- \Rightarrow Each Zero-term yields a quadratic equation in the elements of S .
- \Rightarrow S can be recovered by solving several MQ systems (the hardest with v variables, m equations).

Reconciliation Attack for UOV

Good Keys

$$M_S := \begin{bmatrix} *_{v \times v} & *_{v \times o} \\ *_{o \times v} & *_{o \times o} \end{bmatrix} = \begin{bmatrix} 1_{v \times v} & *_{v \times o} \\ 0_{o \times v} & 1_{o \times o} \end{bmatrix} \begin{bmatrix} *_{v \times v} & 0_{v \times o} \\ *_{o \times v} & *_{o \times o} \end{bmatrix} \quad (1)$$

Only need $M_S = P := \begin{bmatrix} 1_{v \times v} & *_{v \times o} \\ 0_{o \times v} & 1_{o \times o} \end{bmatrix}$

Reconciliation Attack for UOV

Good Keys

Only need $M_S = P := \begin{bmatrix} 1_{v \times v} & *_{v \times o} \\ 0_{o \times v} & 1_{o \times o} \end{bmatrix} = P_{v+1} P_{v+2} \cdots P_n$, and

$$P_n = 1_n + \left[\begin{array}{ccc|c} 0 & \cdots & 0 & \lambda_1 \\ 0 & \cdots & 0 & \lambda_2 \\ \vdots & \ddots & \vdots & \vdots \\ 0 & \cdots & 0 & \lambda_v \\ \hline 0 & \cdots & 0 & 0 \\ \vdots & \ddots & \vdots & \vdots \\ 0 & \cdots & 0 & 0 \end{array} \right]; \quad P_{n-1} = 1_n + \left[\begin{array}{ccc|c|c} 0 & \cdots & 0 & \lambda'_1 & 0 \\ 0 & \cdots & 0 & \lambda'_2 & 0 \\ \vdots & \ddots & \vdots & \vdots & \vdots \\ 0 & \cdots & 0 & \lambda'_v & 0 \\ \hline 0 & \cdots & 0 & 0 & 0 \\ \vdots & \ddots & \vdots & \vdots & \vdots \\ 0 & \cdots & 0 & 0 & 0 \end{array} \right]; \cdots$$

Reconciliation Attack for UOV

Good Keys

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- 1 Perform basis change $w_i := w'_i - \lambda_i w'_n$ for $i = 1 \cdots v$, $w_i = w'_i$ for $i = v + 1 \cdots n$. Evaluate \mathbf{z} in \mathbf{w}' .
- 2 Let all coefficients of $(w'_n)^2$ be zero and solve for the λ_i . We may use any method such as FXL (m equations in $v = v_u = n - o_u$ unknowns).

Reconciliation Attack for UOV

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- 2 Let all coefficients of $(w'_n)^2$ be zero and solve for the λ_i . We may use any method such as FXL (m equations in $v = v_u = n - o_u$ unknowns).
- 3 Set $w'_i := w''_i - \lambda_i w''_{n-1}$ for $i = 1 \cdots v$, and set every $(w''_{n-1})^2$ and $w''_n w''_{n-1}$ term to zero (i.e., more equations in the system). We find P_{n-1} with $2m$ equations in v unknowns. Continue for P_{n-2}, \dots, P_{v+1} .

Summary of UOV

Safe Parameters for UOV(\mathbb{F} , o , v)

security level (bit)	scheme	public key size (kB)	private key size (kB)	hash size (bit)	signature (bit)
80	UOV($\mathbb{F}_{16}, 40, 80$)	144.2	135.2	160	480
	UOV($\mathbb{F}_{256}, 27, 54$)	89.8	86.2	216	648
100	UOV($\mathbb{F}_{16}, 50, 100$)	280.2	260.1	200	600
	UOV($\mathbb{F}_{256}, 34, 68$)	177.8	168.3	272	816
128	UOV($\mathbb{F}_{16}, 64, 128$)	585.1	538.1	256	768
	UOV($\mathbb{F}_{256}, 45, 90$)	409.4	381.8	360	1,080
192	UOV($\mathbb{F}_{16}, 96, 192$)	1,964.3	1,786.7	384	1,152
	UOV($\mathbb{F}_{256}, 69, 138$)	1,464.6	1,344.0	552	1,656
256	UOV($\mathbb{F}_{16}, 128, 256$)	4,644.1	4,200.3	512	1,536
	UOV($\mathbb{F}_{256}, 93, 186$)	3,572.9	3,252.2	744	2,232

What we know today about UOV

- unbroken since 1999 \Rightarrow high confidence in security
- not the fastest multivariate scheme
- very large keys, (comparably) large signatures

Rainbow Digital Signature

Ding and Schmidt, 2004

- Patented by Ding
- May have had patent by T.-T. Moh (expired)
- TTS is its variant with sparse central map

Rainbow Digital Signature

Ding and Schmidt, 2004

- Finite field \mathbb{F} , integers $0 < v_1 < \dots < v_u < v_{u+1} = n$.
- Set $V_i = \{1, \dots, v_i\}$, $O_i = \{v_i + 1, \dots, v_{i+1}\}$, $o_i = v_{i+1} - v_i$.
- Central map Q consists of $m = n - v_1$ polynomials $f^{v_1+1}, \dots, f^{(n)}$ of the form

$$f^{(k)} = \sum_{i,j \in V_\ell} \alpha_{ij}^{(k)} x_i x_j + \sum_{i \in V_\ell, j \in O_\ell} \beta_{ij}^{(k)} x_i x_j + \sum_{i \in V_\ell \cup O_\ell} \gamma_i^{(k)} x_i + \delta^{(k)},$$

with coefficients $\alpha_{ij}^{(k)}$, $\beta_{ij}^{(k)}$, $\gamma_i^{(k)}$ and $\delta^{(k)}$ randomly chosen from \mathbb{F} and ℓ being the only integer such that $k \in O_\ell$.

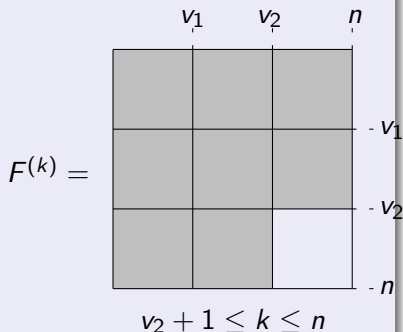
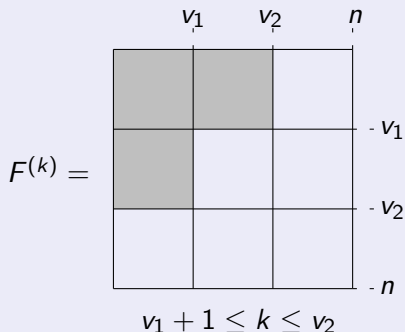
- Choose randomly two affine (or linear) transformations $\mathcal{T} : \mathbb{F}^m \rightarrow \mathbb{F}^m$ and $\mathcal{S} : \mathbb{F}^n \rightarrow \mathbb{F}^n$.
- *public key*: $\mathcal{P} = \mathcal{T} \circ Q \circ \mathcal{S} : \mathbb{F}^n \rightarrow \mathbb{F}^m$
- *private key*: \mathcal{T} , Q , \mathcal{S}

Idea of Rainbow

Inversion of the central map

- Invert the single UOV layers recursively.
- Use the variables of the i -th layer as Vinegars of the $i + 1$ -th layer.

Illustration: Rainbow with two layers



Idea of Rainbow

Inversion of the central map

- Invert the single UOV layers recursively.
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Input: Rainbow central map $Q = (f^{(v_1+1)}, \dots, f^{(n)})$, vector $\mathbf{y} \in \mathbb{F}^m$.

Output: vector $\mathbf{x} \in \mathbb{F}^n$ with $Q(\mathbf{x}) = \mathbf{y}$.

- 1: Choose random values for the variables x_1, \dots, x_{v_1} and substitute these values into the polynomials $f^{(i)}$ ($i = v_1 + 1, \dots, n$).
- 2: **for** $\ell = 1$ to u **do**
- 3: Perform Gaussian Elimination on the polynomials $f^{(i)}$ ($i \in O_\ell$) to get the values of the variables x_i ($i \in O_\ell$).
- 4: Substitute the values of x_i ($i \in O_\ell$) into the polynomials $f^{(i)}$ ($i = v_{\ell+1} + 1, \dots, n$).
- 5: **end for**

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Signature Generation from message d

- 1 Use a hash function $\mathcal{H} : \{0, 1\} \rightarrow \mathbb{F}^m$ to compute $\mathbf{z} = \mathcal{H}(d) \in \mathbb{F}^m$
- 2 Compute $\mathbf{y} = \mathcal{T}^{-1}(\mathbf{z}) \in \mathbb{F}^m$.
- 3 Compute a pre-image $\mathbf{x} \in \mathbb{F}^n$ of \mathbf{y} under the central map \mathcal{Q}
- 4 Compute the signature $\mathbf{w} \in \mathbb{F}^n$ by $\mathbf{w} = \mathcal{S}^{-1}(\mathbf{x})$.

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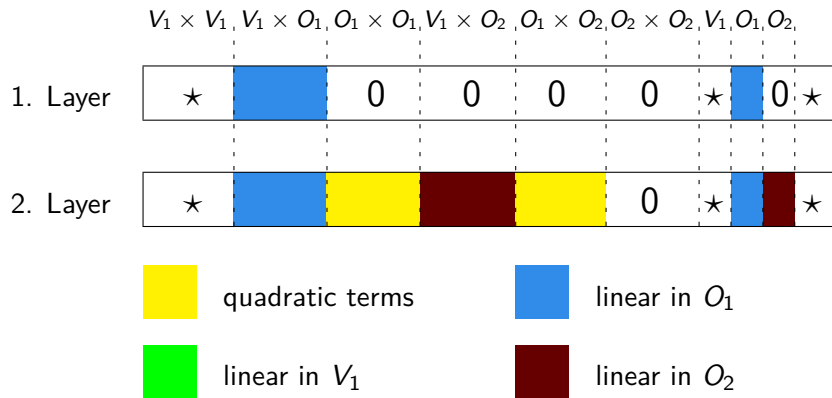
Signature Verification from message d , signature $\mathbf{z} \in \mathbb{F}^n$

- 1 Compute $\mathbf{z} = \mathcal{H}(d)$.
- 2 Compute $\mathbf{z}' = \mathcal{P}(\mathbf{w})$.

Accept the signature $\mathbf{z} \Leftrightarrow \mathbf{w}' = \mathbf{w}$.

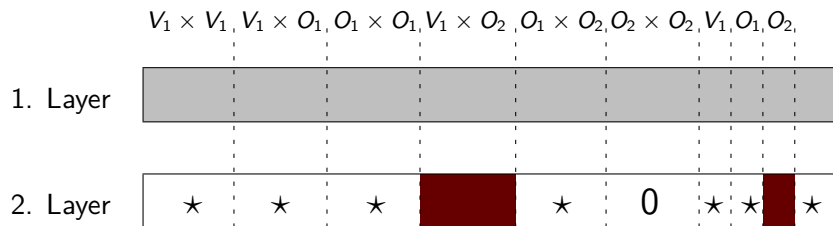
Rainbow Schemes with two layers

Step 1: Choose random values for the Vinegar variables x_1, \dots, x_{V_1} and substitute them into the central polynomials



Rainbow Schemes with two layers

Step 2: Solve the o_1 linear equations given by the polynomials of the first layer for $x_{v_1+1}, \dots, x_{v_2}$ and substitute into the polynomials of the second layer



quadratic terms



linear in O_1



linear in V_1



linear in O_2

Rainbow Schemes with two layers

Step 3: Solve the \mathcal{o}_2 linear equations given by the \mathcal{o}_2 polynomials of the second layer for x_{v_2+1}, \dots, v_n .

Toy Example

- $\mathbb{F} = \text{GF}(7)$, $(v_1, o_1, o_2) = (2, 2, 2)$
- central map $Q = (f^{(3)}, \dots, f^{(6)})$ with

$$f^{(3)} = x_1^2 + 3x_1x_2 + 5x_1x_3 + 6x_1x_4 + 2x_2^2 + 6x_2x_3 + 4x_2x_4 + 2x_2 + 6x_3 + 2x_4 + 5,$$

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

$$f^{(5)} = 2x_1^2 + 3x_1x_2 + 3x_1x_3 + 3x_1x_4 + x_1x_5 + 3x_1x_6 + 6x_1 + 4x_2^2 + x_2x_3 + 4x_2x_4 \\ + x_2x_5 + 3x_2x_6 + 3x_2 + 3x_3x_4 + x_3x_5 + 2x_3x_6 + 2x_3 + 3x_4x_5 + x_5 + 6x_6,$$

$$f^{(6)} = 2x_1^2 + 5x_1x_2 + x_1x_3 + 5x_1x_4 + 5x_1x_6 + 6x_1 + 5x_2^2 + 3x_2x_3 + 5x_2x_5 + 4x_2x_6 \\ + x_2 + 3x_3^2 + 5x_3x_4 + 4x_3x_5 + 2x_3x_6 + 4x_3 + x_4^2 + 6x_4x_5 + 3x_4x_6 \\ + 4x_4 + 4x_5 + x_6 + 2.$$

- Goal: Find pre image $\mathbf{x} \in \mathbb{F}^6$ of $\mathbf{y} = (6, 2, 0, 5)$ under the map Q

Toy Example (2)

- Choose random values for the Vinegar variables x_1 and x_2 , e.g. $(x_1, x_2) = (0, 1)$ and substitute them into the polynomials $f^{(3)}, \dots, f^{(6)}$.

$$\tilde{f}^{(3)} = 5x_3 + 6x_4 + 2, \tilde{f}^{(4)} = x_3 + 5x_4,$$

$$\tilde{f}^{(5)} = 3x_3x_4 + x_3x_5 + 2x_3x_6 + 3x_3 + 3x_4x_5 + 4x_4 + 2x_5 + 2x_6,$$

$$\tilde{f}^{(6)} = 3x_3^2 + 5x_3x_4 + 4x_3x_5 + 2x_3x_6 + x_4^2 + 6x_4x_5 + 3x_4x_6 + 4x_4 + 2x_5 + 5x_6 + 1.$$

- Set $\tilde{f}^{(3)} = y_1 = 6$ and $\tilde{f}^{(4)} = y_2 = 2$ and solve for x_3, x_4
 $\Rightarrow (x_3, x_4) = (3, 4)$
- Substitute into $\tilde{f}^{(5)}$ and $\tilde{f}^{(6)}$
 $\Rightarrow \tilde{\tilde{f}}^{(5)} = 3x_5 + x_6 + 5, \tilde{\tilde{f}}^{(6)} = 3x_5 + 2x_6 + 1$
- Set $\tilde{\tilde{f}}^{(5)} = y_3 = 0$ and $\tilde{\tilde{f}}^{(6)} = y_4 = 5$, solve for x_5 and x_6
 $\Rightarrow (x_5, x_6) = (0, 2)$

A pre image of $\mathbf{y} = (6, 2, 0, 5)$ is given by $\mathbf{x} = (0, 1, 3, 4, 0, 2)$.

Signature Generation

Given: message d

- 1 Use a hash function $\mathcal{H} : \{0, 1\} \rightarrow \mathbb{F}^m$ to compute $\mathbf{w} = \mathcal{H}(d) \in \mathbb{F}^m$
- 2 Compute $\mathbf{x} = \mathcal{S}^{-1}(\mathbf{w}) \in \mathbb{F}^m$.
- 3 Compute a pre-image $\mathbf{y} \in \mathbb{F}^n$ of \mathbf{x} under the central map \mathcal{Q}
- 4 Compute the signature $\mathbf{z} \in \mathbb{F}^n$ by $\mathbf{z} = \mathcal{T}^{-1}(\mathbf{y})$.

Signature Verification

Given: message d , signature $\mathbf{z} \in \mathbb{F}^n$

① Compute $\mathbf{w} = \mathcal{H}(d)$.

② Compute $\mathbf{w}' = \mathcal{P}(\mathbf{z})$.

Accept the signature $\mathbf{z} \Leftrightarrow \mathbf{w}' = \mathbf{w}$.

Security

Rainbow is an extension of UOV

⇒ All attacks against UOV can be used against Rainbow, too.

Additional structure of the central map allows several new attacks

- **MinRank Attack:** Look for linear combinations of the matrices H_i of low rank
- **HighRank Attack:** Look for the linear representation of the variables appearing the lowest number of times in the central polynomials.
- **Rainbow-Band-Separation Attack:** Variant of the UOV-Reconciliation Attack using the additional Rainbow structure [DY08]

Choosing Parameter Selection for Rainbow is interesting

MinRank Attack

Minors Version

Set all rank $r + 1$ minors of $\sum_j \alpha_j H_j$ to 0.

Kernel Vector Guessing Version

- Guess a vector \mathbf{v} , let $\sum_j \alpha_j H_j \mathbf{v} = 0$, hope to find a non-trivial solution.
- (If $m > n$, guess $\lceil \frac{m}{n} \rceil$ vectors.)
- Takes $q^r m^3 / 3$ time to find a r -dimensional subspace.

Accumulation of Kernels and Effective Rank

In the first stage of Rainbow, there are $\alpha_1 = v_2 - v_1$ equations and v_2 variables. The rank should be v_2 . But if your guess corresponds to $x_1 = x_2 = \dots = x_{v_1} = 0$, then about $1/q$ of the time we find a kernel. The easy way to see this is that there are $q^{\alpha_1 - 1}$ different kernels. We say that “effectively the rank is $v_1 + 1$ ”.

Rainbow Band Separation

Extension to UOV reconciliation to use the special Rainbow form.

n variables, $n + m - 1$ quadratic equations

- 1 Let $w_i := w'_i - \lambda_i w'_n$ for $i \leq v$, $w_i = w'_i$ for $i > v$. Evaluate \mathbf{z} in \mathbf{w}' .
- 2 Find m equations by letting all $(w'_n)^2$ terms vanish; there are v of λ_i 's.
- 3 Set all cross-terms involving w'_n in $z_1 - \sigma_1^{(1)} z_{v+1} - \sigma_2^{(1)} z_{v+2} - \cdots - \sigma_o^{(1)} z_m$ to be zero and find $n - 1$ more equations.
- 4 Solve $m + n - 1$ quadratic equations in $o + v = n$ unknowns.
- 5 Repeat, e.g. next set $w''_i := w''_i - \lambda_i w''_{n-1}$ for $i < v$, and let every $(w''_{n-1})^2$ and $w''_n w''_{n-1}$ term be 0. Also set $z_2 - \sigma_1^{(2)} z_{v+1} - \sigma_2^{(2)} z_{v+2} - \cdots - \sigma_o^{(2)} z_m$ to have a zero second-to-last column. [$2m + n - 2$ equations in n unknowns.]

Rainbow - Summary

- no weaknesses found since 2007
- efficient, much faster than RSA
- suitable for low cost devices
- shorter signatures and smaller key sizes than UOV

Parameters

security level (bit)	parameters $\mathbb{F}, v_1, \sigma_1, \sigma_2$	public key size (kB)	private key size (kB)	hash size (bit)	signature (bit)
80	$\mathbb{F}_{16}, 20, 20, 20$	33.4	22.3	160	228
	$\mathbb{F}_{256}, 19, 12, 13$	25.3	19.3	200	352
100	$\mathbb{F}_{16}, 25, 25, 25$	65.9	43.2	200	288
	$\mathbb{F}_{256}, 27, 16, 16$	57.2	44.3	256	472
128	$\mathbb{F}_{16}, 32, 32, 32$	136.6	87.6	256	368
	$\mathbb{F}_{31}, 28, 28, 28$	123.2	74.5	280	420
	$\mathbb{F}_{256}, 36, 21, 22$	136.0	102.5	344	632
192	$\mathbb{F}_{16}, 48, 48, 48$	475.9	301.8	384	564
	$\mathbb{F}_{31}, 44, 40, 40$	360.1	245.2	420	630
256	$\mathbb{F}_{16}, 64, 64, 64$	1,194.4	763.9	512	776

References

- Pa97 J. Patarin: The oil and vinegar signature scheme, presented at the Dagstuhl Workshop on Cryptography (September 97)
- KS98 A. Kipnis, A. Shamir: Cryptanalysis of the Oil and Vinegar Signature scheme. CRYPTO 1998, LNCS vol. 1462, pp. 257–266. Springer, 1988.
- KP99 A. Kipnis, J. Patarin, L. Goubin: Unbalanced Oil and Vinegar Schemes. EUROCRYPT 1999. LNCS vol. 1592, pp. 206–222 Springer, 1999.
- DS05 J. Ding, S. Schmidt: Rainbow, a new multivariate polynomial signature scheme. ACNS 2005. LNCS vol. 3531, pp. 164–175 Springer, 2005.
- DY08 J. Ding, B.Y. Yang, C.H.O. Chen, M.S. Chen, C.M. Cheng: New Differential-Algebraic Attacks and Reparametrization of Rainbow. ACNS 2008, LNCS 5037, pp.242–257, Springer 2008.

Multivariates Part 4: Implementation on Modern CPUs

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Thursday, 28.06.2018

Why are MPKCs Worth Studying?

- Diversification
- Efficiency

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- Diversification: Future-proof against quantum computers.
- Efficiency: Faster than “traditional” PKCs.

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- Diversification: Future-proof against quantum computers.
- Efficiency: Faster than “traditional” PKCs.
... Maybe.

Rate-Determining Mechanisms for MPKCs

Key Generation

Evaluation of coefficients

Public Maps

Evaluating a generic set of quadratic polynomials in $\mathbb{K} = \mathbb{F}_q$

Private Maps

Rate-Determining Mechanisms for MPKCs

Key Generation

Evaluation of coefficients:

- Often as differentials of public map.
- Sometimes, by brute force!

Public Maps

Evaluating a generic set of quadratic polynomials in $\mathbb{K} = \mathbb{F}_q$

Private Maps

Rate-Determining Mechanisms for MPKCs

Key Generation

Evaluation of coefficients

Public Maps

Evaluating a generic set of quadratic polynomials in $\mathbb{K} = \mathbb{F}_q$
usually as a matrix multiplying the vector of monomials

Private Maps

Rate-Determining Mechanisms for MPKCs

Key Generation

Evaluation of coefficients

Public Maps

Evaluating a generic set of quadratic polynomials in $\mathbb{K} = \mathbb{F}_q$

Private Maps

UOV Solving linear systems of equations in $\mathbb{K} = \mathbb{F}_q$

Rainbow Like UOV plus mini “Public Map”

C* High powers in $\mathbb{L} = \mathbb{F}_{q^n}$

HFE Equation solving in $\mathbb{L} = \mathbb{F}_{q^n}$ (general arithmetic)

kHFE Like HFE plus an elimination in \mathbb{L}

Practical Side of Computing

Moore's law

Transistor budget doubles every 18–24 months

Memory Latencies vs Clock Speeds

Year	Hi-End CPU	MHz	DRAM
1979	Z80	2	500ns
1984	80286	10	400ns
1989	80486	40	300ns
1994	Pentium	100	250ns
1999	Athlon	750	200ns
2004	Pentium 4	3800	160ns
2009	Core i7	3200	130ns
2014	Core i7	3400	120ns

Are MPKCs Still Fast?

- Progress in high-precision arithmetic
 - ▶ In 80's, CPUs computed one 32-bit integer product every 15–20 cycles
 - ▶ In 2000, x86 CPUs computed one 64-bit product every 3–10 cycles
 - ▶ Core i7's today produces one 128-bit product every 1 cycle
 - ▶ Marvelous for ECC (and RSA)
- In contrast, progress in \mathbb{F}_{2^q} arithmetic is *slow*
 - ▶ 6502 or 8051: a dozen cycles via three table look-ups
 - ▶ Modern x86: roughly same that many cycles
- Moore's law favors computation, not so much memories
 - ▶ Memory access speed increased at a snail's pace
- Wang et al. made life even harder for MPKCs
 - ▶ Forcing longer message digests
 - ▶ RSA untouched

Questions We Want to Answer

- Can all the extras on modern commodity CPUs help MPKCs as well?
- How have architectural changes affected implementation choices?
- If so, how do MPKCs compare to traditional PKCs today?

SSE, the X86 Vector Instruction Set Extensions

- As packed 8-, 16-, 32- or 64-bit operands
- Move `xmm` to/from `xmm`, memory (even unaligned), x86 registers
- Shuffle data and pack/unpack on vector data
- Bit-wise logical operations like AND, OR, NOT, XOR
- Shift left, right logical/arithmetic by units, or entire `xmm` byte-wise
- Add/subtract on 8-, 16-, 32- and 64-bits
- Multiply 16-bit and 32-bits in various ways
- `VPSHUF`B (32 nibble-to-byte lookup in 1 cycle) and `PALIGNR` (256-bit bitwise rotation) quite powerful

(V)PSHUFb

- “Packed Shuffle Bytes”
 - ▶ Source: (x_0, \dots, x_{15})
 - ▶ Destination: (y_0, \dots, y_{15})
 - ▶ Result: $(y_{x_0 \bmod 32}, \dots, y_{x_{15} \bmod 32})$, treating y_{16}, \dots, y_{31} as 0
- VPSHUFb = two individual PSHUFbS

Speeding Up MPKCs over \mathbb{F}_{16}

- TT : 16×16 table, with $TT_{i,j} = i * j, 0 \leq i, j < 16$
- To compute $a\mathbf{v}$, $a \in \mathbb{F}_{16}, \mathbf{v} \in (\mathbb{F}_{16})^{16}$
 - ▶ $\text{xmm} \leftarrow a$ -th row of TT
 - ▶ $a\mathbf{v} \leftarrow \text{PSHUFB } \text{xmm}, \mathbf{v}$
- Works similarly for $\mathbf{a} \in (\mathbb{F}_{16})^2, \mathbf{v} \in (\mathbb{F}_{16})^{32}$
 - ▶ Need to unpack, do PSHUFBs, then pack
- Delivers $2\times$ performance over simple bit slicing in private map evaluation of rainbow and TTS
- Some other platforms also have similar instructions
 - ▶ AMD's SSE5: PPERM (superset of PSHUFB)
 - ▶ IBM POWER AltiVec/VMX: PERMU
 - ▶ ARM's TBL

Speeding Up MPKCs over \mathbb{F}_{256}

Nibble Slicing

- TL : 256×16 table, with $TL_{i,j} = i * j, 0 \leq i < 256, 0 \leq j < 16$
- TH : 256×16 table, with $TH_{i,j} = i * (16j), 0 \leq i < 256, 0 \leq j < 16$
- To compute $a\mathbf{v}$, $a \in \mathbb{F}_{256}, \mathbf{v} \in (\mathbb{F}_{256})^{16}$
 - ▶ $a\mathbf{v}_i = a(16\lfloor \mathbf{v}_i/16 \rfloor) + a(\mathbf{v}_i \bmod 16), 0 \leq i < 16$
- $\mathbf{v}'_i \leftarrow a(16\lfloor \mathbf{v}_i/16 \rfloor)$
 - ▶ $\mathbf{v}'_i \leftarrow \lfloor \mathbf{v}_i/16 \rfloor$ (SHIFT)
 - ▶ $\text{xmm} \leftarrow a$ -th row of TH
 - ▶ $\mathbf{v}' \leftarrow \text{PSHUFEB xmm}, \mathbf{v}'$
- $\mathbf{v}_i \leftarrow a(\mathbf{v}_i \bmod 16)$
 - ▶ $\mathbf{v}_i \leftarrow \mathbf{v}_i \bmod 16$ (AND)
 - ▶ $\text{xmm} \leftarrow a$ -th row of TL
 - ▶ $\mathbf{v} \leftarrow \text{PSHUFEB xmm}, \mathbf{v}$
- $a\mathbf{v} \leftarrow \mathbf{v} + \mathbf{v}'$ (OR)

Arithmetic in \mathbb{F}_{2^k}

PCLMULQDQ: $\mathbb{F}_2[x]_{<64} \times \mathbb{F}_2[x]_{<64} = \mathbb{F}_2[x]_{<128}$

Of course you use it if you can, sheesh.

Bit-Slicing

Log/Exp Tables to a generator g

Multiplication Tables in Memory (Parallel by VPSHUFb)

Arithmetic in \mathbb{F}_{2^k}

PCLMULQDQ: $\mathbb{F}_2[x]_{<64} \times \mathbb{F}_2[x]_{<64} = \mathbb{F}_2[x]_{<128}$

Bit-Slicing

- Highly parallel — 32/64/128 multiplies at the same time
- Often requires rearranging of data
- Parameters can result in awkward dimensions like $1 + (\text{word size})$
- only good for \mathbb{F}_2 and \mathbb{F}_4 .

Log/Exp Tables to a generator g

Multiplication Tables in Memory (Parallel by VPSHUFb)

Arithmetic in \mathbb{F}_{2^k}

PCLMULQDQ: $\mathbb{F}_2[x]_{<64} \times \mathbb{F}_2[x]_{<64} = \mathbb{F}_2[x]_{<128}$

Bit-Slicing

Log/Exp Tables to a generator g

- Compute xy as $g^{\log_g x + \log_g y}$ if neither is zero.
- $\log 0$ needs to special (-42 for VPSHUFb)
- 3 lookups per mult, some logs can be pre-computed
- Time-constant but method of last choice.

Multiplication Tables in Memory (Parallel by VPSHUFb)

Arithmetic in \mathbb{F}_{2^k}

PCLMULQDQ: $\mathbb{F}_2[x]_{<64} \times \mathbb{F}_2[x]_{<64} = \mathbb{F}_2[x]_{<128}$

Bit-Slicing

Log/Exp Tables to a generator g

Multiplication Tables in Memory (Parallel by VPSHUF B)

- One VPSHUF B per many multiplications in \mathbb{F}_{16}
- How do we do time-constant Table Lookups?

Arithmetic in \mathbb{F}_{2^k}

PCLMULQDQ: $\mathbb{F}_2[x]_{<64} \times \mathbb{F}_2[x]_{<64} = \mathbb{F}_2[x]_{<128}$

Bit-Slicing

Log/Exp Tables to a generator g

Multiplication Tables in Memory (Parallel by VPSHUF B)

Build Multiplication Tables by x_0, x_1, \dots on the Fly:

- 1 Have fixed tables $0 \times, 1 \times, 0x2 \times, 0x3 \times, \dots$
- 2 VPSHUF B to get: $0 \times [x_0, x_1, \dots], 1 \times [x_0, x_1, \dots], 0x2 \times [x_0, x_1, \dots],$
- 3 Transpose, now we can multiply by x_0, x_1, \dots

Some Interesting Design Choices

System and Architecture-Dependent Stuff

- Key Generation
- Matrix-to-Vector-Multiply and Evaluating Public Maps
- Tower Field Arithmetic
- System- and Equation-Solving
 - ▶ Pre-scripted Gröbner Basis Computation
 - ▶ Iterative Methods vs. Gaussian Eliminations
 - ▶ Cantor-Zassenhaus vs. Berlekamp

Key Generation

Matsumoto-Imai's notation: $z_k := \sum_i w_i \left[P_{ik} + Q_{ik} w_i + \sum_{j < i} R_{ijk} w_j \right]$.

Usual Way: as differentials of public map $\mathcal{P} = (p_1, \dots, p_m)$

for $q > 2$, we choose any $a \neq 0, 1$ and get

$$Q_{ik} := (a(a-1))^{-1} (p_k(a\mathbf{v}_i) - ap_k(\mathbf{v}_i))$$

$$P_{ik} := p_k(\mathbf{v}_i) - Q_{ik}$$

$$R_{ijk} := p_k(\mathbf{v}_i + \mathbf{v}_j) - Q_{ik} - Q_{jk} - P_{ik} - P_{jk}$$

For \mathbb{F}_2 , it becomes

$$P_{ik} := p_k(\mathbf{v}_i)$$

$$R_{ijk} := p_k(\mathbf{v}_i + \mathbf{v}_j) - P_{ik} - P_{jk}$$

(\mathbf{v}_i means the unit vector on the i -th direction)

Evaluating Public Maps

Naive Way (and on μP 's still)

$$z_k = \sum_i w_i \left[P_{ik} + Q_{ik} w_i + \sum_{i < j} R_{ijk} w_j \right]$$

For better memory access pattern

- 1 $\mathbf{c} \leftarrow [\mathbf{w}^T, (w_i w_j)_{i < j}]^T$
- 2 $\mathbf{z} \leftarrow \mathbf{P}\mathbf{c}$, where \mathbf{P} is the $m \times n(n+3)/2$ public-key matrix

How to do Matrix-to-Vector mults

Microcontrollers Naively

Somewhat newer CPUs Bit-slicing for \mathbb{F}_{2^k}

With more cache Big look-up tables (with nibble-slicing)

Newest architectures More or less naively, with SSE*

MPKCs over Odd Prime Fields

MPKCs over Odd Prime Fields

Are you out of your mind?

- XOR is easy, addition mod q is not.
- How can it possibly be faster?

MPKCs over Odd Prime Fields

Are you out of your mind?

- XOR is easy, addition mod q is not.
- How can it possibly be faster?

It's more than about speed

- Good for defending against Gröbner basis attacks
 - ▶ The field equation $X^q - X = 0$ becomes much less useful
- SSE* gives you parallel arithmetic on small integers,
 - ▶ and you only need to parallelize 4 or 8 at a time.
- Do you know how many 18-bit multipliers there are on an FPGA?

Basic Building Blocks for Speeding Up Odd MPKCs

PMULHSW

takes upper half in a rounded signed product of two 16-bit words, $\lceil xy/2^{15} \rceil$, good for reduction mod q

VPMADDUSBW

- Packed Multiply and Add, Unsigned and Signed Byte to Word
 - ▶ Source: (x_0, \dots, x_{31}) Unsigned
 - ▶ Destination: (y_0, \dots, y_{31}) Signed
 - ▶ Result: $(x_0y_0 + x_1y_1, x_2y_2 + x_3y_3, \dots, x_{30}y_{30} + x_{31}y_{31})$
- Helpful in evaluating $\mathbf{z} = \mathbf{P}\mathbf{c}$, piece by piece
 - ▶ Let \mathbf{Q} be a 16×2 submatrix of \mathbf{P}
 - ▶ \mathbf{d}^T be the corresponding 2×1 submatrix of \mathbf{c}
 - ▶ $r1 \leftarrow (Q_{11}, Q_{12}, Q_{21}, Q_{22}, \dots, Q_{15,1}, Q_{15,2})$
 - ▶ $r2 \leftarrow (d_1, d_2, d_1, d_2, \dots, d_1, d_2)$
 - ▶ VPMADDUSBW $r1, r2$ computes $\mathbf{Q}\mathbf{d}$
 - ▶ Continue in 16-bits until reduction mod q needed.
- Saves a few mod q operations and delivers $1.5\times$ performance

Big look-up tables for matrix multiplication

As suggested by Berbain *et al*, SAC 2006

- Pre-compute $a\mathbf{v}$ for each column \mathbf{v} in any constant matrix
- Read off the appropriately offset vector as needed
- Can nibble-slice $\mathbb{F}_{16}/\mathbb{F}_{256}$ into $\mathbb{F}_{16}/\mathbb{F}_4$
- Obviously minimizes the need for operations

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Unbelievably ...

Slower than SSE on any modern CPU!

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When L2 isn't fast enough

- SSE instructions have a reverse throughput of 1 cycle today
- memory access is linear when using SSE
- L2 latency 20+ cycles; LUT reads not regular enough
- No way around this today :(

Remarks on Getting More Performance

Laziness often leads to optimality

- Do not always need the tightest range
- The less reductions, the better!
- The less memory access, the better!
- The more regular memory access, the better!
- Packing \mathbb{F}_q -blocks into binary can use more bits than necessary as long as the map is injective and convenient to compute

Wiedemann vs. Gauss Elimination mod q

- How to solve a medium-sized dense linear system?
 - ▶ Wiedemann iterative solver for $\mathbf{Ax} = \mathbf{b}$
 - ★ Compute $\mathbf{zA}^i\mathbf{b}$ for some \mathbf{z}
 - ★ Compute minimal polynomial using Berlekamp-Massey
 - ▶ Requires $O(2n^3)$ field multiplications
 - ▶ Straightforward Gauss elimination requires $O(n^3/3)$
- However, Wiedemann involves much less reductions modulo q
- **However, everything has to be constant-time**
- At the moment Gaussian beats Wiedemann.

To Solve Equation(s) in a Big Tower Field over \mathbb{F}_q

Scripted Gröbner Basis Computation

From 3 quadratic equations in 3 variables, we in succession run Gaussian eliminations on matrices of dimensions 3×10 , 11×19 , 8×16 , 5×13 , with many coefficients that we know to be zero in advance, to reach a degree-8 equation. You can call this a tailored matrix- \mathbf{F}_4 .

Cantor-Zassenhaus (instead of Berlekamp)

- 1 Replace $u(X)$ by $\gcd(u(X), X^{q^k} - X)$ so that u splits in \mathbb{L} .
 - 1 Compute and tabulate $X^d \bmod u(X), \dots, X^{2d-2} \bmod u(X)$.
 - 2 Compute $X^q \bmod u(X)$ via square-and-multiply.
 - 3 Compute and tabulate $X^{q^i} \bmod u(X)$ for $i = 2, 3, \dots, d - 1$.
 - 4 Compute $X^{q^i} \bmod u(X)$ for $i = 2, 3, \dots, k$, then $X^{q^k} \bmod u(X)$.
- 2 Do $\gcd\left(v(X)^{(q^k-1)/2} - 1, u(X)\right)$ for random $v(X)$ with $\deg v < \deg u$, to find nontrivial factor $\geq \frac{1}{2}$ of the time; repeat as needed.

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Scripted Gröbner Basis Computation

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 - 2 Compute $X^q \bmod u(X)$ via square-and-multiply.
 - 3 Compute and tabulate $X^{q^i} \bmod u(X)$ for $i = 2, 3, \dots, d-1$.
 - 4 Compute $X^{q^i} \bmod u(X)$ for $i = 2, 3, \dots, k$, then $X^{q^k} \bmod u(X)$.
- 2 **Toss everything away and repeat unless there is a single solution.**

Anything else New For \mathbb{F}_{2^k} ?

Anything else New For \mathbb{F}_{2^k} ?

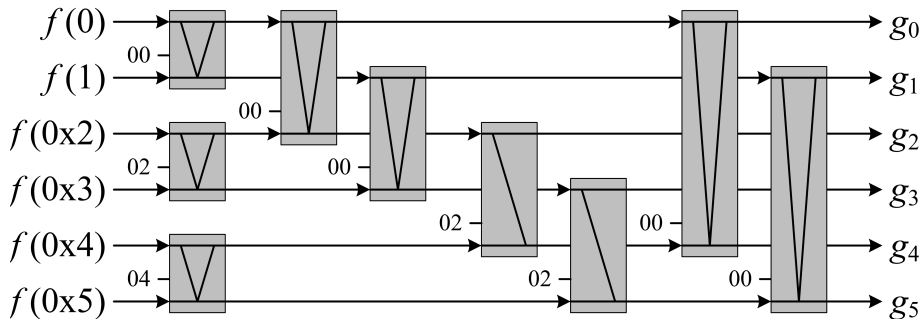
Not Really.

Anything else New For \mathbb{F}_{2^k} ?

Not Really.

Ok, So we implemented some

- Additive-FFT based multiplication using (V)PSHUFb
- TRUNCATED Additive-FFT too



Performance on Xeon E3-1245v3 (Haswell) 3.4GHz

Table: 128-bit MPKCs on Intel Haswell.

schemes	gen-key() M cycles	sign() k cycles	verify() k cycles
Rainbow(16,32,32,32)	154.7	89.9	22.8
Rainbow(31,28,28,28)	93.4	77.4	70.8
Rainbow(256,28,20,20)	581.0	121.6	19.0
PFLASH(16,96-1,64)	78.8	226.0	22.6
GUI(2,240,9,16,16,3)	484.2	4,445.4	197.6
MQDSS-31-64 ^a	1.827	8,510.6	5,752.6
ECDSA(NIST P256) ^b	0.286	377.1	901.5
Ed25519 ^b	0.066	61.0	185.1
RSA-2048 ^b	233.7	5,240.2	66.4
RSA-3072 ^b	844.4	15,400.9	119.3

^a on Core i7-4770K (Haswell) 3.5GHz.

^b eBACS on Xeon E3-1275 v3 (haswell) at 3.5GHz.

Continued: Non-PCLMULQDQ CPUs

We also implemented without PCLMULQDQ, using additive FFT and (V)PSHUFB.

Table: Benchmark of 128-bit MPKCs on SSE-only Platforms

schemes	gen-key() M cycles	sign() k cycles	verify() k cycles
PFLASH(16,96-1,64)	3,269	908.6	32.8
GUI(4,120,17,8,8,2)	510	121,287	1,583.6

Conclusions and Remarks

- It is very important to tune to your architecture.
- MPKCs still competitive speedwise
- Intel's new vector instruction set did double the MPKC throughput.

Thanks for Listening!

- Questions or comments?