



SHANDONG
UNIVERSITY



STRATIVIA®
TECHNOLOGY | CONSULTING | SERVICES

Revisiting the Extension of Matsui's Algorithm 1 to Linear Hulls: Application to TinyJAMBU

Muzhou Li^{1,2} Nicky Mouha³ Ling Sun^{1,2} Meiqin Wang^{1,2,4,✉}

¹Key Laboratory of Cryptologic Technology and Information Security,
Ministry of Education, Shandong University, Jinan, China

²School of Cyber Science and Technology,
Shandong University, Qingdao, China

³Strativia, Largo, MD, USA

⁴Quan Cheng Shandong Laboratory, Jinan, China

March 24, 2023 @ Beijing

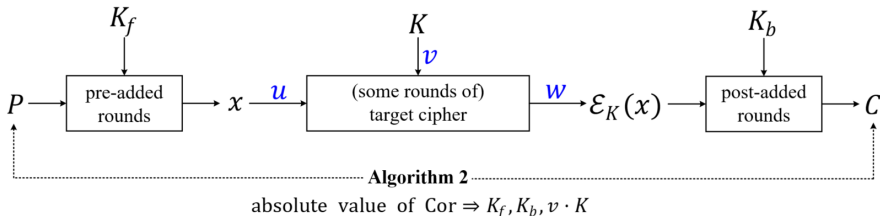
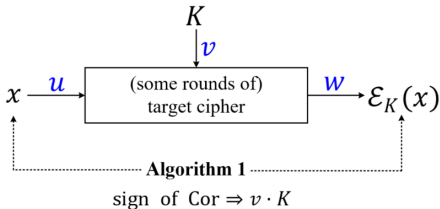
- 1 Motivation and Contribution
- 2 Previous Extension of Matsui's Algorithm 1
- 3 New Methodology and Statistical Models
- 4 Application to TinyJAMBU
- 5 Conclusion and Future Work

- 1 Motivation and Contribution
- 2 Previous Extension of Matsui's Algorithm 1
- 3 New Methodology and Statistical Models
- 4 Application to TinyJAMBU
- 5 Conclusion and Future Work

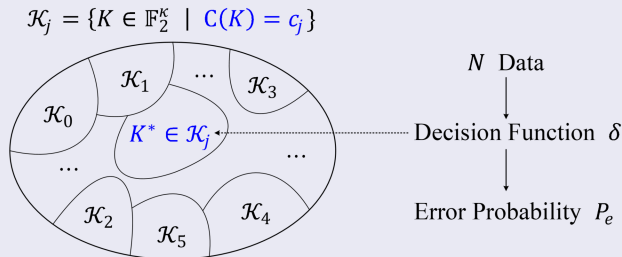
Motivation

Linear Cryptanalysis [Matsui @ EUROCRYPT 1993]

$$u \cdot x \oplus w \cdot \mathcal{E}_K(x) = v \cdot K \text{ (high correlation Cor)}$$



Linear Hull Version of Algorithm 1 [Röck & Nyberg @ DCC]



Relation between N and P_e is **not accurately described** (Experiments)

- ▲ **Inaccuracy** comes from the **methodology** of deducing this relation.
- ▲ Algorithm 1 is more suitable than Algorithm 2: for ciphers where **only part of the state can be obtained**.

New Statistical Models

Absolute Error $\max |P_e^{\text{theory}} - P_e^{\text{expr.}}| =$

1. Previous Methodology \Rightarrow 93.75% (MLE)
2. Our Methodology \Rightarrow 1.9% \searrow (MLE), 2.19% \searrow (Threshold)

Key Recovery Attacks on TinyJAMBU [Wu & Huang]

	Method	Attack Phase	Attacked/ Total	Key Len. Supported	Key Bits Rec.	Reference
v1	Cube	Ini. & Enc.	2604/3200	128	1-bit	Teng et al. @ ePrint2021/1164
	Linear Hull	Tag Gen.	384/384	all	$\geq 7\text{-bit}$	Our
v2	Linear Hull	Tag Gen.	387/640	all	$\geq 7\text{-bit}$	Our

First cryptanalysis results in the **nonce-respecting** setting on the **full TinyJAMBU v1** and the **round-reduced TinyJAMBU v2**.

New Statistical Models

Absolute Error $\max |P_e^{\text{theory}} - P_e^{\text{expr.}}| =$

1. Previous Methodology \Rightarrow 93.75% (MLE)
2. Our Methodology \Rightarrow 1.9% \searrow (MLE), 2.19% \searrow (Threshold)

Key Recovery Attacks on TinyJAMBU [Wu & Huang]

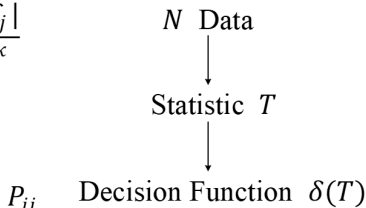
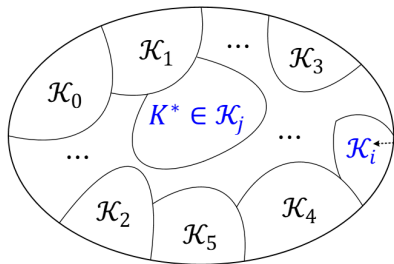
	Method	Attack Phase	Attacked/ Total	Key Len. Supported	Key Bits Rec.	Reference
v1	Cube	Ini. & Enc.	2604/3200	128	1-bit	Teng et al. @ ePrint2021/1164
	Linear Hull	Tag Gen.	384/384	all	\geq 7-bit	Our
v2	Linear Hull	Tag Gen.	387/640	all	\geq 7-bit	Our

First cryptanalysis results in the **nonce-respecting** setting on the **full TinyJAMBU v1** and the **round-reduced TinyJAMBU v2**.

- 1 Motivation and Contribution
- 2 Previous Extension of Matsui's Algorithm 1
- 3 New Methodology and Statistical Models
- 4 Application to TinyJAMBU
- 5 Conclusion and Future Work

Previous Extension: Key Recovery Framework

$$\mathcal{K}_j = \{K \in \mathbb{F}_2^\kappa \mid \mathcal{C}(K) = c_j\}, \quad \pi_j = \frac{|\mathcal{K}_j|}{2^\kappa}$$



$$P_j^e = \sum_{i \neq j} P_{ij}$$

$$P_e = \sum_j \pi_j P_j^e \quad (\text{average over all keys})$$

► Relation between N and P_e should be depicted accurately. ◀

Previous Extension: Methodology

Given desired P_e :

- $P_{ij} = \frac{P_e}{m-1}$ (**assumption**) with m being the number of all key classes
- N_{ij} : data needed in making decision between \mathcal{K}_i and \mathcal{K}_j
- Construct the relation between P_{ij} and N_{ij} with statistical models
- $N = \max N_{ij}$ (**upper bound**)

Such methodology causes the **inaccuracy**.

Previous Extension: Three Attack Settings

Direct Attack:

$$\mathcal{K}_i = \{K \in \mathbb{F}_2^K \mid C(K) = c_i\}$$

Basic RK: fixed key difference α

$$\mathcal{K}_i^\alpha = \{K \in \mathbb{F}_2^K \mid C(K) - C(K \oplus \alpha) = c_i\}$$

Multiple RK: t differences $\alpha_0, \alpha_1, \dots, \alpha_{t-1}$ (form a basis)

- Proceed basic rk attack **under each α_j** , and obtain **guessed $\mathcal{K}_{\eta_j}^{\alpha_j}$**
- If all these attacks succeed, K^* must belong to $\bigcap_{0 \leq j \leq t-1} \mathcal{K}_{\eta_j}^{\alpha_j}$

Key Information Obtained

Direct Attack < Basic RK < Multiple RK

- 1 Motivation and Contribution
- 2 Previous Extension of Matsui's Algorithm 1
- 3 New Methodology and Statistical Models**
- 4 Application to TinyJAMBU
- 5 Conclusion and Future Work

Threshold-Based Statistical Model (Direct Attack)

New Methodology

For each j :

- Depict clearly the **distribution \mathcal{D}_j** of T related to N when $K^* \in \mathcal{K}_j$
- Compute P_{ij} with the **CDF of \mathcal{D}_j** and $\delta(T) = i$

Get $P_e = \sum_j \pi_j \sum_{i \neq j} P_{ij}$

Statistic & Distribution

Let N_0^K records how many x fulfill the linear hull given N known data x .

$$T = 2 \frac{N_0^K}{N} - 1 \sim \mathcal{D}_j = \mathcal{N} \left(c_j, \sigma^2 = \frac{1 - c_j^2}{N} B \right)$$

when $K^* \in \mathcal{K}_j$, where $B = 1$ (KP sampling) or $\frac{2^n - N}{2^n - 1}$ (DKP sampling).

Threshold-Based Statistical Model (Direct Attack)

New Methodology

For each j :

- Depict clearly the **distribution \mathcal{D}_j** of T related to N when $K^* \in \mathcal{K}_j$
- Compute P_{ij} with the **CDF of \mathcal{D}_j** and $\delta(T) = i$

Get $P_e = \sum_j \pi_j \sum_{i \neq j} P_{ij}$

Statistic & Distribution

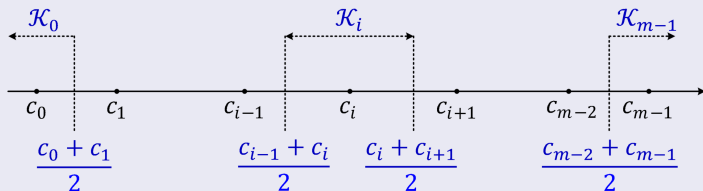
Let N_0^K records how many x fulfill the linear hull given N known data x .

$$T = 2 \frac{N_0^K}{N} - 1 \sim \mathcal{D}_j = \mathcal{N} \left(c_j, \sigma^2 = \frac{1 - c_j^2}{N} B \right)$$

when $K^* \in \mathcal{K}_j$, where $B = 1$ (KP sampling) or $\frac{2^n - N}{2^n - 1}$ (DKP sampling).

Threshold-Based Statistical Model (Direct Attack)

Threshold-Based Decision Function $\delta(T)$

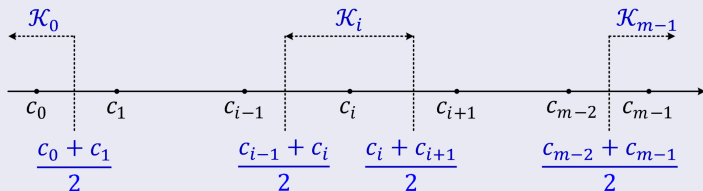


$$\begin{aligned} P_{ij} &= \Pr[\delta(T) = i | T \sim D_j] \\ &= \Pr \left[\frac{c_{i-1} + c_i}{2} < T \leq \frac{c_i + c_{i+1}}{2} \mid T \sim \mathcal{N}(c_j, \sigma^2) \right] \\ &= \Phi \left(\frac{c_i + c_{i+1}}{2} - c_j \right) - \Phi \left(\frac{c_{i-1} + c_i}{2} - c_j \right) \end{aligned}$$

where $\Phi(\cdot)$ denotes the CDF of $\mathcal{N}(0, 1)$.

Threshold-Based Statistical Model (Direct Attack)

Threshold-Based Decision Function $\delta(T)$



$$\begin{aligned} P_{ij} &= \Pr[\delta(T) = i | T \sim D_j] \\ &= \Pr \left[\frac{c_{i-1} + c_i}{2} < T \leq \frac{c_i + c_{i+1}}{2} \mid T \sim \mathcal{N}(c_j, \sigma^2) \right] \\ &= \Phi \left(\frac{c_i + c_{i+1}}{2} - c_j \right) - \Phi \left(\frac{c_{i-1} + c_i}{2} - c_j \right) \end{aligned}$$

where $\Phi(\cdot)$ denotes the CDF of $\mathcal{N}(0, 1)$.

Experimental Verification on Threshold-Based Models

Use 256-round keyed permutation of TinyJAMBU

Linear Hull: $\{7, 30, 37, 44, 54, 64, 77, 81, 84, 91, 98, 118, 121\} \rightarrow \{64\}$

- Trail 1: 7, 27, 30, 37, 44, 81, 111, 118 (Cor = $+2^{-10}$)
- Trail 2: 6, 7, 27, 30, 37, 44, 81, 111, 118 (Cor = -2^{-11})
- Trail 3: 7, 21, 27, 30, 37, 44, 81, 111, 118 (Cor = -2^{-11})
- Trail 4: 6, 7, 21, 27, 30, 37, 44, 81, 111, 118 (Cor = $+2^{-11}$)

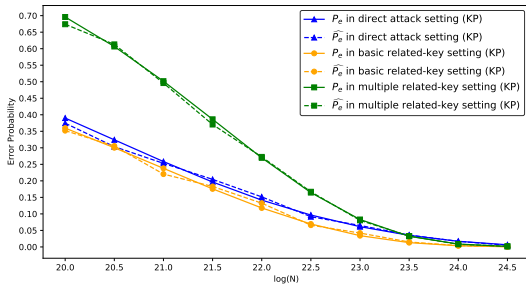
Let $ek = k_7 \oplus k_{27} \oplus k_{30} \oplus k_{37} \oplus k_{44} \oplus k_{81} \oplus k_{111} \oplus k_{118}$. The whole key space $k_{21} || ek || k_6 \in \mathbb{F}_2^3$ is divided into four disjoint classes:

$$\mathcal{K}(c_0 = -2.5 \cdot 2^{-10}) = \{7\}, \mathcal{K}(c_1 = -0.5 \cdot 2^{-10}) = \{2, 3, 6\},$$

$$\mathcal{K}(c_2 = +0.5 \cdot 2^{-10}) = \{0, 1, 4\}, \mathcal{K}(c_3 = +2.5 \cdot 2^{-10}) = \{5\}.$$

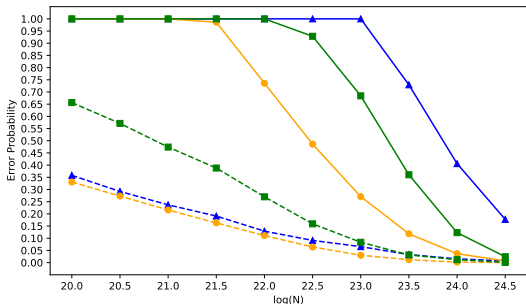
Key information can be recovered using our statistical models.

Experimental Verification on Threshold-Based Models



Our Threshold-Based

2.19%



Previous MLE-Based

93.45%

Direct Attack (KP Sampling)

- $\delta(T) = i \iff ML(T, i) > ML(T, t)$ for $\forall t \neq i$
- For each t :
 - When $p_i > p_t$,

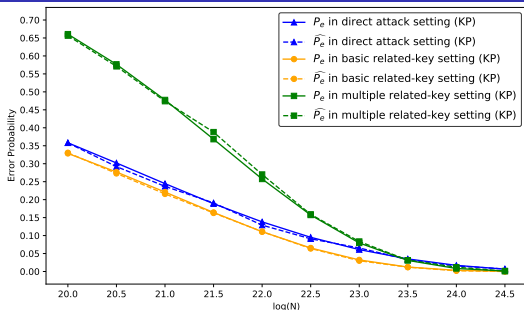
$$\frac{\log_2 \pi_t - \log_2 \pi_i + N \log_2(1 - p_t) - N \log_2(1 - p_i)}{\log_2 p_i - \log_2(1 - p_i) - \log_2 p_t + \log_2(1 - p_t)} < T < N$$

- When $p_i < p_t$,

$$0 < T < \frac{\log_2 \pi_t - \log_2 \pi_i + N \log_2(1 - p_t) - N \log_2(1 - p_i)}{\log_2 p_i - \log_2(1 - p_i) - \log_2 p_t + \log_2(1 - p_t)}$$

- Intersection of above $(m - 1)$ intervals: $N_{\min}^i < T < N_{\max}^i$
- $P_{ij} = \Phi_{N, p_j}^b(N_{\max}^i) - \Phi_{N, p_j}^b(N_{\min}^i)$. Φ_{N, p_j}^b is the CDF of T .

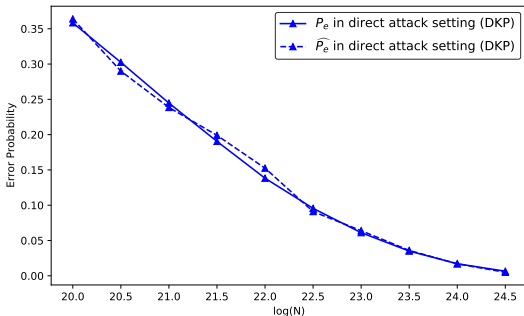
Experimental Verification on MLE-Based Models



Our MLE-Based

1.9%

slightly more precise
but
much slower to compute

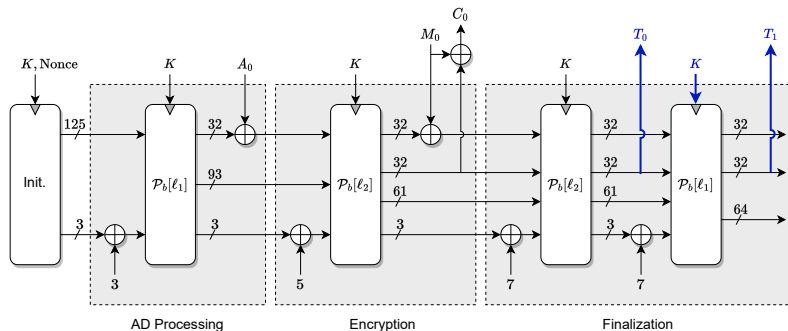


- 1 Motivation and Contribution
- 2 Previous Extension of Matsui's Algorithm 1
- 3 New Methodology and Statistical Models
- 4 Application to TinyJAMBU**
- 5 Conclusion and Future Work

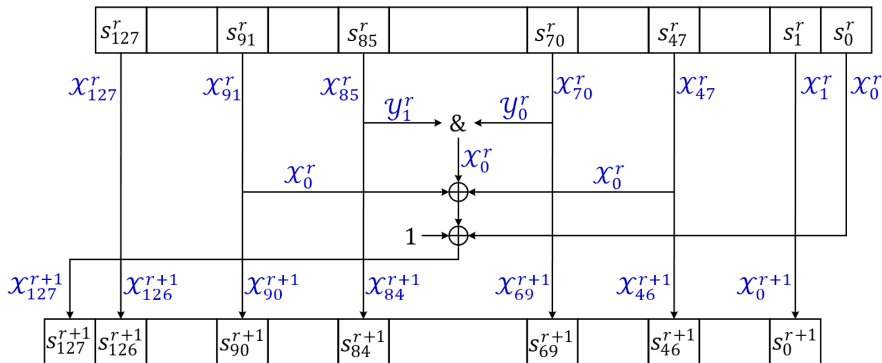
Brief Introduction to TinyJAMBU

TinyJAMBU [Wu & Huang]:

- Key: 128-, 192-, 256-bit
- $\mathcal{P}_b : (s_{127}, s_{126}, \dots, s_0) \rightarrow (z, s_{127}, \dots, s_1)$
$$z = s_0 \oplus s_{47} \oplus (\sim (s_{70} \& s_{85})) \oplus s_{91} \oplus k_i$$
- $l_1 = 384$ (v1); $l_1 = 640$ (v2)



Correlation Evaluation of Linear Trail



For R-round trail:

$$\lambda_0 \cdot T_0 \oplus \lambda_1 \cdot T_1 = \mathcal{X}^0 \cdot x^0 \oplus \mathcal{X}^R \cdot x^R = \bigoplus_{r=0}^{R-1} (\mathcal{X}^r \cdot x^r \oplus \mathcal{X}^{r+1} \cdot x^{r+1}) = \bigoplus_{s=0}^{14} f_s$$

Correlation Evaluation of Linear Trail

$$\begin{aligned}
 f_s = & \chi_0^s (\chi_{70}^s \& \chi_{85}^s) \oplus \chi_0^{s+15} (\chi_{85}^s \& \chi_{85}^{s+15}) \oplus \chi_0^{s+30} (\chi_{85}^{s+15} \& \chi_{85}^{s+30}) \\
 & \oplus \dots \oplus \chi_0^{s+15t_s} (\chi_{85}^{s+15(t_s-1)} \& \chi_{85}^{s+15t_s}) \\
 & \oplus \chi_0^s \gamma_0^s \chi_{70}^s \\
 & \oplus \left[\chi_0^s \gamma_1^s \oplus \chi_0^{s+15} \gamma_0^{s+15} \right] \chi_{85}^s \\
 & \oplus \left[\chi_0^{s+15} \gamma_1^{s+15} \oplus \chi_0^{s+30} \gamma_0^{s+30} \right] \chi_{85}^{s+15} \oplus \dots \\
 & \oplus \left[\chi_0^{s+15(t_s-1)} \gamma_1^{s+15(t_s-1)} \oplus \chi_0^{s+15t_s} \gamma_0^{s+15t_s} \right] \chi_{85}^{s+15(t_s-1)} \\
 & \oplus \chi_0^{s+15t_s} \gamma_1^{s+15t_s} \chi_{85}^{s+15t_s} \\
 & \oplus \bigoplus_{j=0}^{t_s} \chi_0^{s+15j} (1 \oplus k_{s+15j \bmod \kappa})
 \end{aligned}$$

f_s contains several Boolean functions with **chained AND gates**.

Correlation evaluation of the linear trail \iff evaluate these f_s

Correlation Evaluation of Linear Trail

$$f(x_0, \dots, x_n) = x_0 \& x_1 \oplus x_1 \& x_2 \oplus \dots \oplus x_{n-1} \& x_n \oplus a_0 x_0 \oplus \dots \oplus a_n x_n$$

Absolute Correlation [Song et al.]

1. n Odd: $|\text{Cor}(f)| = 2^{-(n+1)/2}$.
2. n Even: $|\text{Cor}(f)| = 2^{-n/2}$ if $\bigoplus_{j=0}^{n/2} a_{2j} = 0$; otherwise, $|\text{Cor}(f)| = 0$.

$$\text{Sign}(f) = \prod_{i=0}^{t-1} (-1)^{\left(\bigoplus_{j=0}^i a_{2j}\right) a_{2i+1}}, t = \begin{cases} \frac{n+1}{2}, & n \text{ Odd} \\ \frac{n}{2}, & n \text{ Even and } \bigoplus_{j=0}^t a_{2j} = 0 \end{cases}$$

Key Bits Involved

$$\lambda_0 \cdot T_0 \oplus \lambda_1 \cdot T_1 \approx \bigoplus_{r=0}^{R-1} \mathcal{X}_0^r (1 \oplus k_r \bmod \kappa)$$

k_j is involved in the trail if $\bigoplus_{r \in \mathcal{J}} \mathcal{X}_0^r = 1$, $\mathcal{J} = \{r \mid j = r \bmod k\}$

Correlation Evaluation of Linear Trail

$$f(x_0, \dots, x_n) = x_0 \& x_1 \oplus x_1 \& x_2 \oplus \dots \oplus x_{n-1} \& x_n \oplus a_0 x_0 \oplus \dots \oplus a_n x_n$$

Absolute Correlation [Song et al.]

1. n Odd: $|\text{Cor}(f)| = 2^{-(n+1)/2}$.
2. n Even: $|\text{Cor}(f)| = 2^{-n/2}$ if $\bigoplus_{j=0}^{n/2} a_{2j} = 0$; otherwise, $|\text{Cor}(f)| = 0$.

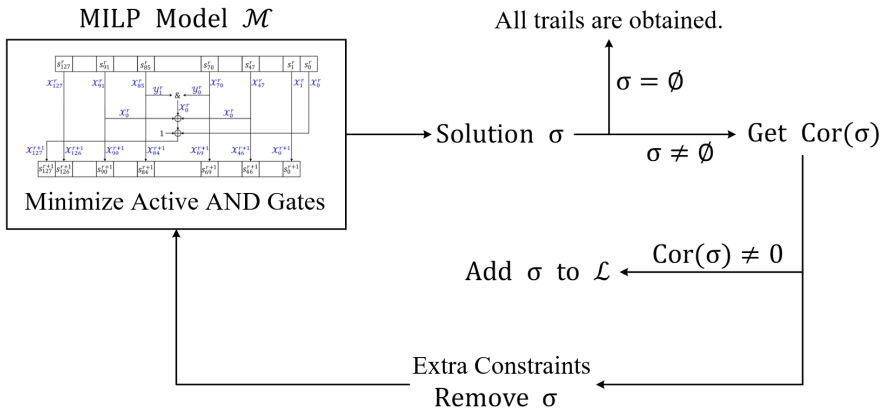
$$\text{Sign}(f) = \prod_{i=0}^{t-1} (-1)^{\left(\bigoplus_{j=0}^i a_{2j}\right) a_{2i+1}}, t = \begin{cases} \frac{n+1}{2}, & n \text{ Odd} \\ \frac{n}{2}, & n \text{ Even and } \bigoplus_{j=0}^t a_{2j} = 0 \end{cases}$$

Key Bits Involved

$$\lambda_0 \cdot T_0 \oplus \lambda_1 \cdot T_1 \approx \bigoplus_{r=0}^{R-1} \mathcal{X}_0^r (1 \oplus k_{r \bmod \kappa})$$

k_j is involved in the trail if $\bigoplus_{r \in \mathcal{J}} \mathcal{X}_0^r = 1$, $\mathcal{J} = \{r \mid j = r \bmod \kappa\}$

Searching All Linear Trails in a Given Hull for TinyJAMBU



Linear Hull used in Attacking V1

384-Round Linear Hull ($\lambda_0 = 0x8024C000, \lambda_1 = 0x00220808$):

Cor.	$+2^{-42}$	$+2^{-43}$	$+2^{-44}$	$+2^{-45}$	$+2^{-46}$	$+2^{-47}$	$+2^{-48}$
Trail	1	10	39	92	120	81	82
Cor.	-2^{-42}	-2^{-43}	-2^{-44}	-2^{-45}	-2^{-46}	-2^{-47}	-2^{-48}
Trail	2	11	36	93	117	82	84

850 trails are composed in this hull.

Key Recovery Attacks on Full V1 (Threshold-Based)

Setting	Type	N	Users	Key Info. Rec.	\Pr_{success}	$\kappa = 128$	$\kappa \in \{192, 256\}$
Direct Attack	weak-key multi-user	$2^{96.8}$	2^{50}	7.639	82.35 %		✓
	single-user	$2^{96.8}$	1	7.639	82.35 %	✓	✓
Basic RK	weak-key multi-user	$2^{97.1}$	2^{50}	8.033	86.16 %		✓
	single-user	$2^{97.1}$	1	8.033	86.16 %	✓	✓
Multiple RK	weak-key multi-user	$2^{102.31}$	2^{56}	14.063	84.85 %		✓
	single-user	$2^{102.31}$	1	14.063	84.85 %	✓	✓

Data limits: the number of tags collected per key should $\leq 2^{47}$.

Weak-key multi-user: each user has their own key but with some bits in common.

- 1 Motivation and Contribution
- 2 Previous Extension of Matsui's Algorithm 1
- 3 New Methodology and Statistical Models
- 4 Application to TinyJAMBU
- 5 Conclusion and Future Work**

Conclusion and Future Work

Conclusion: New Statistical Models

Absolute Error $\max |P_e^{\text{theory}} - P_e^{\text{expr.}}|$:

- Threshold-based: 2.19% & MLE-based: 1.9%
- Röck and Nyberg: 93.45%

Improvements on accuracy are due to our new methodology.

Conclusion: Cryptanalysis of TinyJAMBU

- Full v1 & Round-Reduced v2

Partial key bits are recovered in the nonce-respecting setting.

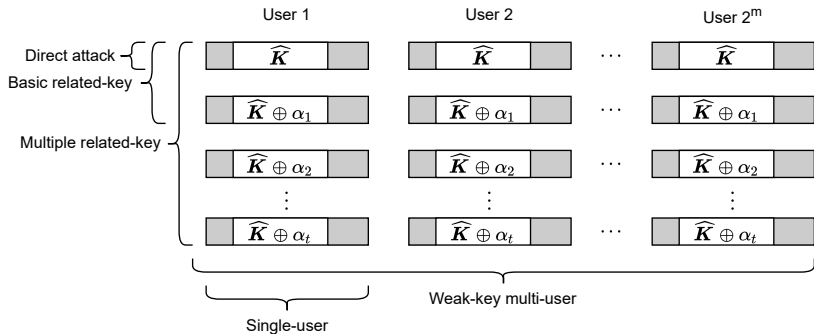
Future Work

- Further applications of our models
- Investigate whether different masks can recover more key bits

Thanks for Your Attention!
Any Questions?

Backup Slides

Attack Settings



Security margin of TinyJAMBU in the multi-user setting will drop from 2^d to 2^{d-m} when 2^m different values of keys are used.