

Round-Based Approximation of (Higher-Order) Differential-Linear Correlation

A Geometric Approach Perspective

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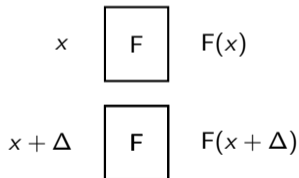
Background Knowledge

Differential-linear attacks [LH94]

- Differential-linear distinguisher for $F : \mathbb{F}_2^n \rightarrow \mathbb{F}_2^n$:

$$DL[\Delta \xrightarrow{F} \lambda] = \text{Cor}[\lambda^\top (F(x) + F(x + \Delta))] = \frac{1}{2^n} \sum_{x \in \mathbb{F}_2^n} (-1)^{v^\top D_\Delta F(x)}$$

where $D_\Delta F(x) = F(x) + F(x + \Delta)$



Question: How to approximate $DL[\Delta \xrightarrow{F} \lambda]$?

Existing approximation methods (1)

- Classical 2-phase approximation [LH94]: $F = F_1 \circ F_0$, $D[\Delta \xrightarrow{F_0} \Delta'] = p$, $\text{Cor}[\lambda' \xrightarrow{F_1} \lambda] = \varepsilon$

$$\text{DL}[\Delta \xrightarrow{F} \lambda] = \pm p\varepsilon^2$$

Two assumptions:

- 1 Independence between F_0 and F_1
 - 2 When $\Delta \xrightarrow{F_0} \Delta'$ does **NOT** happen with $1 - p$ probability, the correlation is 0
- Classical 3-phase refinement [DSK08]: $F = F_2 \circ F_1 \circ F_0$, $D[\Delta \xrightarrow{F_0} \Delta'] = p$, $\text{Cor}[\lambda' \xrightarrow{F_2} \lambda] = \varepsilon$
Experiments for F_1 : $\text{Cor}[\Delta' \xrightarrow{F_1} \lambda'] = c$

$$\text{DL}[\Delta \xrightarrow{F} \lambda] = pc\varepsilon^2$$

- Closed formula, removing Assumption 2 [BLN14]: $F = F_1 \circ F_0$

$$\text{DL}[\Delta \xrightarrow{F} \lambda] = \sum_{\lambda'} \text{DL}[\Delta \xrightarrow{F_0} \lambda'] \text{Cor}[\lambda' \xrightarrow{F_1} \lambda]$$

Existing approximation methods (2)

- Differential-linear connectivity table (DLCT) [BDK+19]:

$$F = F_2 \circ F_1 \circ F_0, \quad D[\Delta \xrightarrow{F_0} \Delta'] = p, \quad \text{Cor}[\lambda' \xrightarrow{F_2} \lambda] = \varepsilon$$

$$\text{DLCT for } F_1 \text{ (1 layer of S-boxes): } \text{Cor}[\Delta' \xrightarrow{F_1} \lambda'] = \text{DLCT}[\Delta', \lambda']$$

$$\text{DL}[\Delta \xrightarrow{F} \lambda] = p \cdot \text{DLCT}[\Delta', \lambda'] \cdot \varepsilon^2$$

- Algebraic transition form (ATF) [LLL21]: denote $f(x) := \lambda^\top F(x)$

$$\lambda^\top (F(x) + F(x + \Delta)) = f(x) + f(x + \Delta) = \sum_{u \in \mathbb{F}_2} f(x + u \cdot \Delta) = \text{Coe}_f(u)$$

$$\text{DL}[\Delta \xrightarrow{F} \lambda] := \text{Cor}[\lambda^\top (F(x) + F(x + \Delta))] = \text{Cor}[\text{Coe}_f(u)]$$

F is iterative: $F = F_{r-1} \circ F_{r-2} \circ \dots \circ F_0$

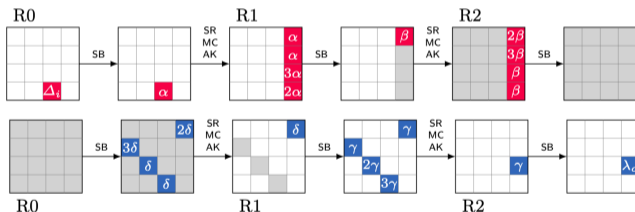
$$x + u\Delta \xrightarrow{F_0} \alpha_1 + u\beta_1 \xrightarrow{F_1} \dots \xrightarrow{F_{r-1}} \alpha_r + u\beta_r$$

Approximate the bias of bits of α_i and β_i according to [ANFs]

Existing approximation methods (3)

- Generalized differential-linear connectivity table (DLCT) [HDE+24]:
Using a boomerang way, different tables are constructed to **cover more middle rounds**

$$DL[\Delta_i \rightarrow \lambda_o] = \sum_{\alpha, \beta, \gamma, \delta} \text{Cor}_{UDLCT}(\Delta_i, \alpha, \delta) \cdot \text{Cor}_{EDLCT}(\alpha, \beta, \delta, \gamma) \cdot \text{Cor}_{LDLCT}(\beta, \gamma, \lambda_o)$$



(The picture was copied from [HDE+24])

- Truncated differential table (TDT) [PZW+24]: trace the propagation of **truncated** differentials

A meta-approach that enables systematically thinking about various cryptanalytic methods

- Free vector space: Regarding elements in \mathbb{F}_2^n as **basis vectors**, where u is denoted by δ_u . We can construct a **vector space**:

$$\mathbb{R}[\mathbb{F}_2^n] = \left\{ \sum_u k_u \delta_u, k_u \in \mathbb{R}, u \in \mathbb{F}_2^n \right\} \quad (\mathbb{R} \text{ is the field of real numbers})$$

- Obtain a linear extension function (pushforward) from any (nonlinear) function
For $F : \mathbb{F}_2^n \rightarrow \mathbb{F}_2^n$, we define T^F as

$$T^F : \mathbb{R}[\mathbb{F}_2^n] \rightarrow \mathbb{R}[\mathbb{F}_2^n]; \quad \sum_u k_u \delta_u \mapsto \sum_u k_u \delta_{F(u)}$$

Nonlinear F can be embedded into a much larger **linear map** T^F

Transition matrix and change-of-basis [Bey21]

- $T^F : \mathbb{R}[\mathbb{F}_2^n] \rightarrow \mathbb{R}[\mathbb{F}_2^n]$ is a linear map. Fixing bases for the input/output spaces, we will get a matrix **w.r.t the bases**
- Regard $(\delta_u, u \in \mathbb{F}_2^n)$ as the standard basis for the input and output spaces, the corresponding **transition matrix** has elements as

$$T^F[v, u] = \delta_v^T T^F(\delta_u) = \delta_v(F(u)) \quad \left(\delta_u(x) := \begin{cases} 1, & u = x \\ 0, & \text{otherwise} \end{cases} \right)$$

- Choose another basis $(\beta_u, u \in \mathbb{F}_2^n)$ satisfying $(\delta_u, u \in \mathbb{F}_2^n) = (\beta_u, u \in \mathbb{F}_2^n) \cdot H$

$$C^F = H T^F H^{-1}$$

When H is the **Walsh-Hadamard matrix**, $(\beta_u, u \in \mathbb{F}_2^n)$ is called the **linear basis**. C^F is called the **correlation matrix** [DGV94].

- The coordinate of C^F is

$$C^F[v, u] = 2^{-n} \sum_{x \in \mathbb{F}_2^n} (-1)^{u^T x + v^T F(x)}$$

d -wise form of a function

Definition (d -wise form of a function [HZC+25])

The d -wise form of a function $F : \mathbb{F}_2^n \rightarrow \mathbb{F}_2^n$ is defined as

$$\begin{aligned} \mathbf{F}^{\times d} : \mathbb{F}_2^{n \times d} &\rightarrow \mathbb{F}_2^{n \times d} \\ (x, \theta_1, \theta_2, \dots, \theta_{d-1}) &\mapsto (F(x), D_{\theta_1}F(x), \dots, D_{\theta_{d-1}}F(x)) \end{aligned}$$

$D_{\theta}F(x)$ is the derivative of F in the direction of θ at the point x , i.e., $D_{\theta}F(x) = F(x) + F(x + \theta)$.

Correlation matrix of $\mathbf{F}^{\times d}$:

$$\mathbf{C}^{\mathbf{F}^{\times d}} = \left(\bigotimes_{i=0}^{d-1} \mathbf{H} \right) \cdot \mathbf{T}^{\mathbf{F}^{\times d}} \cdot \left(\bigotimes_{i=0}^{d-1} \mathbf{H} \right)^{-1} = \left(\bigotimes_{i=0}^{d-1} \mathbf{H} \right) \cdot \mathbf{T}^{\mathbf{F}^{\times d}} \cdot \left(\bigotimes_{i=0}^{d-1} \mathbf{H}^{-1} \right)$$

\otimes : Kronecker/tensor product

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$$C^{F^{\times d}} = \left(\bigotimes_{i=0}^{d-1} H \right) \cdot T^{F^{\times d}} \cdot \left(\bigotimes_{i=0}^{d-1} H \right)^{-1} = \left(\bigotimes_{i=0}^{d-1} H \right) \cdot T^{F^{\times d}} \cdot \left(\bigotimes_{i=0}^{d-1} H^{-1} \right)$$

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Remark. The d -wise form was called d -th order form [HZC+25]. But “order” has been used in higher-order differential attacks, so we change its name since this work.

Properties of correlation matrix

Theorem ([DCV94, Bey21])

The correlation matrix has the following properties

(1) For $F = F_{r-1} \circ F_{r-2} \circ \dots \circ F_0$

$$C^F = \prod_{i=0}^{r-1} C^{F_i}$$

(2) For $F = \underbrace{f \parallel f \parallel \dots \parallel f}_t$

$$C^F = \bigotimes_{i=0}^{t-1} C^f$$

A Geometric Approach Viewpoint on DL

DL correlation and correlation vector

Definition (Correlation vector of a multiset)


The **correlation vector** of a multiset \mathbb{S} whose values are taken from \mathbb{F}_2^n is a vector $\text{CV}(\mathbb{S}) \in \mathbb{R}^{2^n}$ whose u -th coordinate is

$$\text{CV}(\mathbb{S})[u] = \frac{1}{|\mathbb{S}|} \sum_{x \in \mathbb{S}} (-1)^{u^\top x}$$

Remark: There are similar things before defined for Boolean functions

New perspective on DL correlation of $F : \mathbb{F}_2^n \rightarrow \mathbb{F}_2^n$ from the **geometric approach**:

- Consider $F^{\times 2} : (x, \Delta) \mapsto (F(x), D_\Delta F(x))$
- Input multiset: $\mathbb{P} = \mathbb{F}_2^n \times \{\Delta\}$, output multiset: $\mathbb{C} := F^{\times 2}(\mathbb{P})$
- $\text{DL}[\Delta \xrightarrow{F} \lambda]$ is the **$(0, \lambda)$ -coordinate of $\text{CV}(\mathbb{C})$** :

$$\text{CV}(\mathbb{C})[(0, \lambda)] = 2^{-n} \sum_{F(x), D_\Delta F(x)} (-1)^{(0, \lambda)^\top (F(x), D_\Delta F(x))} = 2^{-n} \sum_{x \in \mathbb{F}_2^n, \Delta \in \{\Delta\}} (-1)^{\lambda^\top D_\Delta F(x)} = \text{DL}[\Delta \xrightarrow{F} \lambda]$$


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- **Question:** how to **approximate** $\text{CV}(\mathbb{C})[(0, \lambda)]$?

How to approximate $\text{CV}(\mathbb{C})[(0, \lambda)]$?

Lemma (Correlation vector propagation, [DVG94, Bey21])

Consider $F: \mathbb{F}_2^n \rightarrow \mathbb{F}_2^n$, and $\mathbb{C} = F(\mathbb{P})$

$$\text{CV}(\mathbb{C}) = C^F \cdot \text{CV}(\mathbb{P})$$

Similarly for $F^{\times 2}$: $\text{CV}(\mathbb{C}') = C^{F^{\times 2}} \cdot \text{CV}(\mathbb{P}')$.

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$F^{\times 2}$ is composite

$$F^{\times 2} = F_{r-1}^{\times 2} \circ F_{r-2}^{\times 2} \circ \cdots \circ F_0^{\times 2}$$

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According to the correlation matrix/geometric approach theories,

$$C^{F^{\times 2}} = C^{F_{r-1}^{\times 2}} \cdot C^{F_{r-2}^{\times 2}} \dots C^{F_0^{\times 2}}$$

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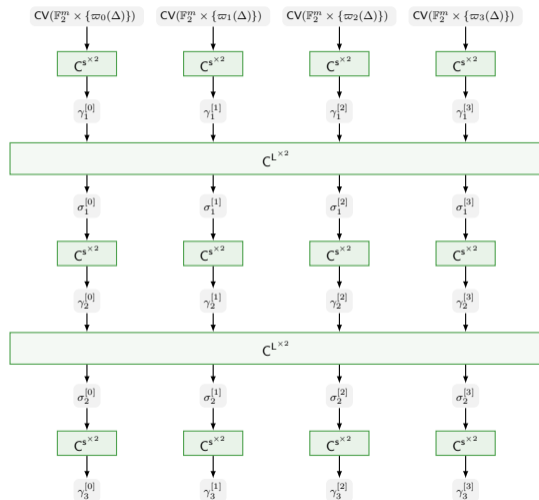
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Visualization for the round-based approximation



Global setting:

- **S**-box size: m
- **block** size: $n = mt$

Prepare for the plaintext vector σ_0

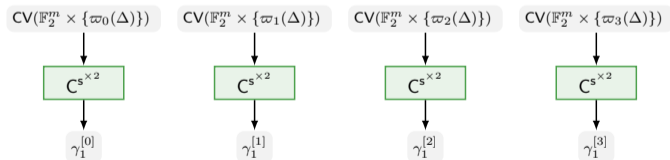
- The plaintext multiset: $\mathbb{P} = \mathbb{F}_2^n \times \{\Delta\} = \prod_{i=0}^{t-1} (\mathbb{F}_2^m \times \varpi_i(\Delta))$.

Projection map: For $v = (v_0, v_1, \dots, v_{t-1}) \in \mathbb{F}_2^{mt}$, $\varpi_i : \mathbb{F}_2^{mt} \rightarrow \mathbb{F}_2^m$; $v \mapsto v_i$.

- Compute the correlation vector

$$\text{CV}(\mathbb{P}) = \bigotimes_{i=0}^{t-1} \text{CV}(\mathbb{F}_2^m \times \varpi_i(\Delta))$$

- Complexity: $\mathcal{O}(t \times 2^m)$



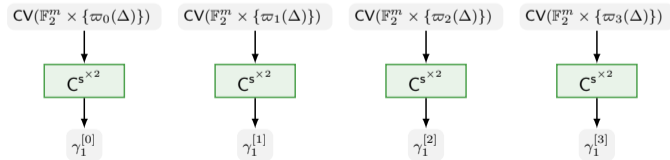
Approximation for the first nonlinear layer $\mathbf{S} = \underbrace{\mathbf{s} \|\mathbf{s}\| \cdots \|\mathbf{s}\| \mathbf{s}}_t$

- Compute the first nonlinear layer:

$$\gamma_1 = \mathbb{C}^{s \times 2} \cdot \text{CV}(\mathbb{P}) = \mathbb{C}^{s \times 2} \sigma_0 := \mathbb{C}^{s \times 2} \cdot \bigotimes_{i=0}^{t-1} \sigma_0^{[i]} = \bigotimes_{i=0}^{t-1} \mathbb{C}^{s \times 2} \cdot \bigotimes_{i=0}^{t-1} \sigma_0^{[i]} = \bigotimes_{i=0}^{t-1} \mathbb{C}^{s \times 2} \cdot \sigma_0^{[i]}$$

Raw complexity: $\mathcal{O}(t \times 2^{4m})$

- A fast algorithm can reduce the time complexity to $\mathcal{O}(t \times 2^{3m})$



Approximation for the first linear layer: L

- **Cannot** handle the linear layer efficiently:

$$\sigma_1 = C^{L \times 2} \cdot \gamma_1 = C^{L \times 2} \cdot \bigotimes_{i=0}^{t-1} \gamma_1^{[i]}$$

Because $C^{L \times 2}$ cannot be decomposed into small matrices aligned to S-boxes

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- **However**, $\sigma_1[(v_0, v_1)] = \gamma_1[(L^{L \times 2})^T (v_0, v_1)] = \gamma_1[(L^T v_0, L^T v_1)]$

$$= \prod_{j=0}^{t-1} \gamma_1^{[j]}[(\varpi_j(L^T v_0), \varpi_j(L^T v_1))]$$

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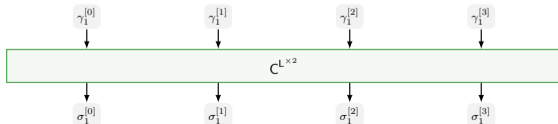
$$= \prod_{j=0}^{t-1} \gamma_1^{[j]}[(\varpi_j(L^T v_0), \varpi_j(L^T v_1))]$$

- Compute coordinates corresponding with S-boxes

$$\mathbb{I} = \left\{ \sigma_1[\tau_i(w)] : w \in \mathbb{F}_2^{2m}, 0 \leq i < t \right\}$$

Inclusion map: $\tau_i : \mathbb{F}_2^m \rightarrow \mathbb{F}_2^{mt}; \quad v \mapsto (0, \dots, v, \dots, 0)$

- Complexity: $\mathcal{O}(t \times 2^{2m})$



What about the remaining coordinates of σ_1 ?

Assumption (Piling-up assumption)

For any $v \in \mathbb{F}_2^{2m \times t}$,

$$\varphi[v] \approx \prod_{i=0}^{t-1} \varphi[\tau_i(\varpi_i(v))]$$

Given any coordinate $v = (v_0, v_1, \dots, v_{t-1})$

$$\sigma_1[(v_0, v_1, \dots, v_{t-1})] \approx \prod_{i=0}^{t-1} \sigma_1^{[i]}[v_i]$$

Do not need to store the whole σ_1 !

What about the remaining coordinates of σ_1 ?

Assumption (Piling-up assumption)

For any $v \in \mathbb{F}_2^{2m \times t}$,

$$\varphi[v] \approx \prod_{i=0}^{t-1} \varphi[\tau_i(\varpi_i(v))]$$

Given any coordinate $v = (v_0, v_1, \dots, v_{t-1})$

$$\sigma_1[(v_0, v_1, \dots, v_{t-1})] \approx \prod_{i=0}^{t-1} \sigma_1^{[i]}[v_i]$$

Do not need to store the whole σ_1 !

Remark. This assumption holds if we assume that the S-boxes in a round are all independent.

Influence of key-XORs: **fixed-key** setting

- k is a constant, and the 2-wise form of the key-XOR K is

$$K_k^{\times 2} : \mathbb{F}_2^n \times \mathbb{F}_2^n \rightarrow \mathbb{F}_2^n; \quad (x, \theta) \rightarrow (x + k, \theta)$$

- The **correlation matrix** of $K_k^{\times 2}$ has the element

$$\begin{aligned} C^{K_k^{\times 2}}[(v_0, v_1)(u_0, u_1)] &= 2^{-2n} \sum_{x \in \mathbb{F}_2^n, \theta \in \mathbb{F}_2^n} (-1)^{u_0^\top x + u_1^\top \theta + v_0^\top (x+k) + v_1^\top \theta} \\ &= (-1)^{v_0^\top k} \left(2^{-n} \sum_{x \in \mathbb{F}_2^n} (-1)^{(u_0^\top + v_0^\top)x} \right) \left(2^{-n} \sum_{\theta \in \mathbb{F}_2^n} (-1)^{(u_1^\top + v_1^\top)\theta} \right) \\ &= (-1)^{v_0^\top k} \delta(u_0 + v_0) \delta(u_1 + v_1). \end{aligned}$$

- Let $\varphi = C^{K_k^{\times 2}} \sigma$, then

$$\varphi[(v_0, v_1)] = \sum_{u_0 \in \mathbb{F}_2^n, u_1 \in \mathbb{F}_2^n} (-1)^{v_0^\top k} \delta(v_0 + u_0) \delta(v_1 + u_1) \sigma[(u_0, u_1)] = (-1)^{v_0^\top k} \sigma[(v_0, v_1)].$$

Conclusion: the key values only influence the sign

Influence of key-XORs: average-key setting

- k is a uniform and independent input

$$\mathbf{K}^{\times 2} : \mathbb{F}_2^n \times \mathbb{F}_2^n \times \mathbb{F}_2^n \rightarrow \mathbb{F}_2^n; (x, \theta, k) \rightarrow (x + k, \theta).$$

- The correlation matrix of \mathbf{K} has the elements as

$$\begin{aligned} \mathbf{C}^{\mathbf{K}^{\times 2}}[(v_0, v_1), (u_0, u_1, u_k)] &= 2^{-3n} \sum_{x \in \mathbb{F}_2^n, \theta \in \mathbb{F}_2^n, k \in \mathbb{F}_2^n} (-1)^{u_0^\top x + u_1^\top \theta + u_k^\top k + v_0^\top (x+k) + v_1^\top \theta} \\ &= \delta(v_0 + u_k) \delta(v_0 + u_0) \delta(v_1 + u_1). \end{aligned}$$

- Let $\varphi = \mathbf{C}^{\mathbf{K}^{\times 2}} \sigma$. σ' and σ'' are the correlation vectors of the data path and key, respectively

$$\varphi[(v_0, v_1)] = \sigma'[(v_0, v_1)] \cdot \sigma''[v_0].$$

Under the assumption that k is uniform,

$$\sigma''[v_0] = 2^{-n} \sum_{k \in \mathbb{F}_2^n} (-1)^{v_0^\top k} = \begin{cases} 1, & v_0 = 0, \\ 0, & v_0 \neq 0. \end{cases}$$

Consequently,

$$\varphi[(v_0, v_1)] = \begin{cases} 0, & v_0 \neq 0, \\ \sigma'[(v_0, v_1)], & v_0 = 0. \end{cases}$$

Conclusion: the key has no influence

DL Application Results to S-box Ciphers

Application to Ascon

Target	Rounds	Exp. Cor.	Th. Cor.	Method	Ref.
Ascon-128 init.	4	2^{-1}	2^{-5}	DLCT	[BDK+19]
			$2^{-1.36}$	ATF	[LLL, C21]
			$2^{-1.09}$	HATF	[HPT+23]
			2^{-1}	TDT	[PZW+24]
			2^{-1}	Round-based	[Ours]
Ascon-128	5	$2^{-8.94}$	$2^{-9.1}$	TDT	[PZW+24]
			$2^{-8.94}$	Round-based	[Ours]
			$2^{-7.94}$	Round-based	[Ours]
Ascon-128a init.	6 ★	–	$2^{-23.89}$	Round-based	[Ours]
Ascon- p	5	$2^{-4.33}$	$2^{-4.0}$	GDLCT	[HDE24]
			$2^{-4.21}$	Round-based	[Ours]
			$2^{-7.61}$	$2^{-6.83}$	GDLCT
			$2^{-7.58}$	Round-based	[Ours]
	6 ★		$2^{-21.89}$	Round-based	[Ours]

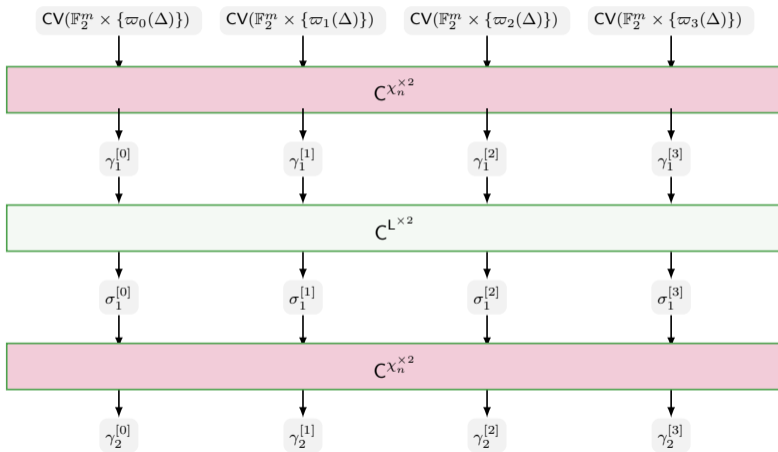
★ means we find DL distinguishers for **more rounds**

Application to Present

Target	Rounds	Exp. Cor.	Th. Cor.	Method	Ref.
Present	10	$2^{-12.94}$	$2^{-13.97}$ $2^{-13.03}$	GDLCT Round-based	[HDE24] [Ours]
	13		$2^{-27.01}$ $2^{-22.43}$	GDLCT Round-based	[HDE24] [Ours]
	17 ★		$2^{-28.85}$	Round-based	[Ours]
	18 ★		$2^{-31.46}$	Round-based	[Ours]

Round-Based Approximation for χ -Based Ciphers

Visualization for the round-based approximation for Chi ciphers



Chi, Chi^{×2} and their correlations

- Chi function [Dae95]

$$\chi_n : \mathbb{F}_2^n \rightarrow \mathbb{F}_2^n; \quad (x[0], x[1], \dots, x[n-1]) \mapsto (y[0], y[1], \dots, y[n-1]), \quad y[i] = x[i] + (x[i+1] + 1)x[i+2]$$

- Auxiliary function g_n

$$g_n : \mathbb{F}_2^n \rightarrow \mathbb{F}_2^n; \quad (x[0], x[1], \dots, x[n-1]) \mapsto (y[0], y[1], \dots, y[n-1]), \quad y[i] = (x[i] + 1)x[i+1]$$

Chi, $\text{Chi}^{\times 2}$ and their correlations

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- Auxiliary function g_n

$$g_n : \mathbb{F}_2^n \rightarrow \mathbb{F}_2^n; \quad (x[0], x[1], \dots, x[n-1]) \mapsto (y[0], y[1], \dots, y[n-1]), \quad y[i] = (x[i] + 1)x[i+1]$$

- $\chi_n(x) = x + (g_n(x) \lll 1)$

$$C^{\chi_n}[v, u] := \text{Cor}[u^\top x + v^\top \chi_n(x)]$$

Chi, $\text{Chi}^{\times 2}$ and their correlations

- Chi function [Dae95]

$$\chi_n : \mathbb{F}_2^n \rightarrow \mathbb{F}_2^n; \quad (x[0], x[1], \dots, x[n-1]) \mapsto (y[0], y[1], \dots, y[n-1]), \quad y[i] = x[i] + (x[i+1] + 1)x[i+2]$$

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$$g_n : \mathbb{F}_2^n \rightarrow \mathbb{F}_2^n; \quad (x[0], x[1], \dots, x[n-1]) \mapsto (y[0], y[1], \dots, y[n-1]), \quad y[i] = (x[i] + 1)x[i+1]$$

- $\chi_n(x) = x + (g_n(x) \lll 1)$

$$C^{\chi_n}[v, u] := \text{Cor}[u^\top x + v^\top \chi_n(x)] = \text{Cor}[(u^\top + v^\top)x + (v \ggg 1)^\top g_n(x)] = C^{g_n}[v \ggg 1, v + u]$$

Chi, Chi^{×2} and their correlations

- Chi function [Dae95]

$$\chi_n : \mathbb{F}_2^n \rightarrow \mathbb{F}_2^n; \quad (x[0], x[1], \dots, x[n-1]) \mapsto (y[0], y[1], \dots, y[n-1]), \quad y[i] = x[i] + (x[i+1] + 1)x[i+2]$$

- Auxiliary function g_n

$$g_n : \mathbb{F}_2^n \rightarrow \mathbb{F}_2^n; \quad (x[0], x[1], \dots, x[n-1]) \mapsto (y[0], y[1], \dots, y[n-1]), \quad y[i] = (x[i] + 1)x[i+1]$$

- $\chi_n(x) = x + (g_n(x) \lll 1)$

$$C^{\chi_n}[v, u] := \text{Cor}[u^\top x + v^\top \chi_n(x)] = \text{Cor}[(u^\top + v^\top)x + (v \ggg 1)^\top g_n(x)] = C^{g_n}[v \ggg 1, v + u]$$

- The coordinate of $C^{\chi_n^{\times 2}}$ can be computed by

$$\begin{aligned} C^{\chi_n^{\times 2}}[(v_0, v_1), (u_0, u_1)] &= C^{\chi_n}[v_0, u_0] \cdot C^{\chi_n}[v_0 + v_1, u_0 + u_1] \quad ([HZC + 25]) \\ &= C^{g_n}[v_0 \ggg 1, v_0 + u_0] \cdot C^{g_n}[(v_0 + v_1) \ggg 1, v_0 + v_1 + u_0 + u_1] \end{aligned}$$

- To compute $\gamma = C^{\chi_n^{\times 2}}\sigma$, we need $C^{g_n}[v, u]$

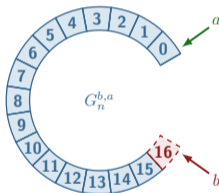
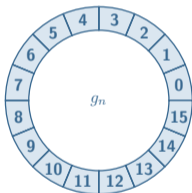
Break the circle of g_n

Definition ($G_n^{a,b}$)

We define

$$G_n^{b,a} : \mathbb{F}_2^n \rightarrow \mathbb{F}_2^n; \quad y = G_n(x)$$

where $y[i] = (x[i] + 1)x[i + 1]$, $0 \leq i < n$ and $x[0] = a, x[n] = b$.



Proposition (Correlation of $C^{G_n^{b,a}}$)

$$C^{g_n}[v, u] = C^{G_n^{0,0}}[v, u] + C^{G_n^{1,1}}[v, u].$$

Chaining property of $C_n^{a,b}$ coordinates

Proposition

Let $n > 1$, $v'' = v \parallel v'$, $u'' = u \parallel u' \in \mathbb{F}_2^{n+1}$ where $u', v' \in \mathbb{F}_2$. From $C_n^{b,a}[v, u]$ we can get $C_n^{b,a}[v'', u'']$ from the following formula,

$$C_{n+1}^{b,a}[v'', u''] = \frac{1}{2} \sum_{c \in \mathbb{F}_2} (-1)^{u'c + v'(c+1)b} C_n^{c,a}[v, u]$$

Lemma

For the four possible values of $v', u', a, b \in \mathbb{F}_2$, we define a matrix $M_{v',u'}$ whose (b, a) -coordinate is $M_{v',u'}[b, a] = \frac{1}{2}(-1)^{ua+v(a+1)b}$. According to the four possibilities of (v', u') , we obtain four matrices:

$$M_{0,0} = \frac{1}{2} \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}, M_{0,1} = \frac{1}{2} \begin{bmatrix} 1 & -1 \\ 1 & -1 \end{bmatrix}, M_{1,0} = \frac{1}{2} \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}, M_{1,1} = \frac{1}{2} \begin{bmatrix} 1 & -1 \\ -1 & -1 \end{bmatrix}$$

$C_n^{b,a}[v, u]$ can be computed as

$$C_n^{b,a}[v, u] = e_b^\top \left(\prod_{i=0}^{n-1} M_{v[i],u[i]} \right) e_a$$

where $e_0 = [1, 0]^\top$ and $e_1 = [0, 1]^\top$

Theorem

Denote $e_0 = [1, 0]^T$ and $e_1 = [0, 1]^T$. The coordinate of $C^{\chi^{\times 2}}$ can be computed as

$$C^{\chi_n^{\times 2}}[(v_0, v_1), (u_0, u_1)] = \sum_{i=0}^1 \sum_{i'=0}^1 (e_i \otimes e_{i'})^T \prod_{j=0}^{n-1} (M_{v_0, v_1, u_0, u_1, j}^{\otimes}) (e_i \otimes e_{i'}),$$

where $M_{v_0, v_1, u_0, u_1, j}^{\otimes} = M_{((v_0+v_1) \ggg 1)[j], (u_0+u_1+v_0+v_1)[j]} \otimes M_{(v_1 \ggg 1)[j], (u_1+v_1)[j]}$.

Approximate $\gamma[(v_0, v_1)]$

- Let $\gamma = \mathbf{C}^{x_n \times 2} \sigma$. Assume $\mathbb{J} = \{\sigma[\tau_i(w)] : w \in \mathbb{F}_2^2, i = 0, \dots, n-1\}$ is known, we want to compute

$$\mathbb{I} = \left\{ \gamma[\tau_i(w)] : w \in \mathbb{F}_2^2, i = 0, \dots, n-1 \right\} \text{ (a bit-level approximation)}$$

Approximate $\gamma[(v_0, v_1)]$

- Let $\gamma = \mathbb{C}^{X_n^{X_2}} \sigma$. Assume $\mathbb{J} = \{\sigma[\tau_i(w)] : w \in \mathbb{F}_2^2, i = 0, \dots, n-1\}$ is known, we want to compute

$$\mathbb{I} = \left\{ \gamma[\tau_i(w)] : w \in \mathbb{F}_2^2, i = 0, \dots, n-1 \right\} \text{ (a bit-level approximation)}$$

- Assume $\sigma[(u_0, u_1)] = \prod_{i=0}^{t-1} \sigma^{[i]}[(\varpi_i(u_0), \varpi(u_1))]$

$$\begin{aligned} \gamma[(v_0, v_1)] &= \sum_{(u_0, u_1) \in \mathbb{F}_2^{2n}} \mathbb{C}^{X_2^{X_2}} [(v_0, v_1), (u_0, u_1)] \cdot \sigma[(u_0, u_1)] \\ &= \sum_{(u_0, u_1) \in \mathbb{F}_2^{2n}} \sum_{i=0}^1 \sum_{i'=0}^1 (e_i \otimes e_{i'})^\top \prod_{j=0}^{n-1} M_{v_0, v_1, u_0, u_1, j}^\otimes \cdot (e_i \otimes e_{i'}) \cdot \sigma[(u_0, u_1)] \\ &= \sum_{i=0}^1 \sum_{i'=0}^1 (e_i \otimes e_{i'})^\top \sum_{(u_0, u_1) \in \mathbb{F}_2^{2n}} \prod_{j=0}^{n-1} M_{v_0, v_1, u_0, u_1, j}^\otimes \prod_{j=0}^{n-1} \sigma^{[j]}[(\varpi_j(u_0), \varpi(u_1))] \cdot (e_i \otimes e_{i'}) \\ &= \sum_{i=0}^1 \sum_{i'=0}^1 (e_i \otimes e_{i'})^\top \sum_{(u_0, u_1) \in \mathbb{F}_2^{2n}} \prod_{j=0}^{n-1} (\sigma^{[j]}[(\varpi_j(u_0), \varpi_j(u_1))] M_{v_0, v_1, u_0, u_1, j}^\otimes) \cdot (e_i \otimes e_{i'}) \\ &= \sum_{i=0}^1 \sum_{i'=0}^1 (e_i \otimes e_{i'})^\top \prod_{j=0}^{n-1} \sum_{u_0[j], u_1[j] \in \mathbb{F}_2} (\sigma^{[j]}[\varpi_j((u_0, u_1))] M_{v_0, v_1, u_0, u_1, j}^\otimes) \cdot (e_i \otimes e_{i'}). \end{aligned}$$

DL Application Results to Chi Ciphers

Application to Subterranean-2.0 and Koala- p

Target	Rounds	Exp. Cor.	Th. Cor.	Method	Ref.
Subterranean-2.0	5 ★	$-2^{-7.88}$	$-2^{-7.98}$	Round-based	[Ours]
	6 ★		$2^{-20.09}$	Round-based	[Ours]
Koala- p	5 ★	$2^{-6.73}$	$2^{-6.87}$	Round-based	[Ours]
	6 ★		$2^{-21.42}$	Round-based	[Ours]

Higher-Order Differential-Linear Attacks

Round-based approximation for 2^d -wise form

- d -th Higher-order differential linear (HDL) attacks

$$\text{HDL}[\Delta \xrightarrow{F} \lambda] = 2^{-n} \sum_{x \in \mathbb{F}_2^n} (-1)^{\lambda \cdot \sum_{\mathbf{w} \in \mathbb{F}_2^d \setminus \{0\}} D_{\langle \mathbf{w}, \Delta \rangle} F(x)}$$

$\Delta = (\Delta_0, \Delta_1, \dots, \Delta_{d-1})$ is d linearly-independent vectors

- Higher-order algebraic transition form [HPT+23] is the only method to approximate a non-deterministic HDL correlation
- Let $\mathbf{v} = (0, \underbrace{\lambda, \dots, \lambda}_{2^d - 1 \text{ times}}) \in \mathbb{F}_2^{n \times 2^d}$

$$\begin{aligned} \text{CV}(\mathbb{C})[\mathbf{v}] &= \sum_{\mathbf{u}} C^{F^{2^d}}[\mathbf{v}, \mathbf{u}] \cdot \text{CV}(\mathbb{P})[\mathbf{u}] = 2^{-n} \sum_{x \in \mathbb{F}_2^n} (-1)^{\lambda \cdot \sum_{\mathbf{u} \in \mathbb{F}_2^d \setminus \{0\}} D_{\langle \mathbf{u}, \Delta \rangle} F(x)} \\ &= \text{CV}(\mathbb{C})[0, \underbrace{\lambda, \dots, \lambda}_{2^d - 1 \text{ times}}] \\ &= \text{HDL}[\Delta \xrightarrow{F} \lambda] \end{aligned}$$

- Round-based approximation

$$\text{CV}(\mathbb{P}) = \sigma_0 \xrightarrow{C^{S \times 2^d}} \gamma_1 \xrightarrow{C^{L \times 2^d}} \sigma_1 \rightarrow \dots \rightarrow \sigma_{r-1} \xrightarrow{C^{S \times 2^d}} \gamma_r = \text{CV}(\mathbb{C}).$$

Correlation propagation for $\chi_n^{\times 2^d}$

$$\begin{aligned}
 \gamma[\mathbf{v}] &= \sum_{\mathbf{u}} C^{\chi_n^{\times 2^d}}[\mathbf{v}, \mathbf{u}] \cdot \sigma[\mathbf{u}] \\
 &= \sum_{i^{(0)}=0}^1 \cdots \sum_{i^{(2^d-1)}=0}^1 \left(\bigotimes_{j=0}^{2^d-1} \mathbf{e}_{i^{(j)}}^\top \right) \left(\sum_{u_0, \dots, u_{2^d-1} \in \mathbb{F}_2^n} \prod_{j=0}^{n-1} \sigma[\tau_j(\varpi_j(\mathbf{u}))] \bigotimes_{k=0}^{2^d-1} M_{\lambda_0^{(k)}[j], \lambda_1^{(k)}[j]} \right) \left(\bigotimes_{j=0}^{2^d-1} \mathbf{e}_{i^{(j)}} \right) \\
 &= \sum_{i^{(0)}=0}^1 \cdots \sum_{i^{(2^d-1)}=0}^1 \left(\bigotimes_{j=0}^{2^d-1} \mathbf{e}_{i^{(j)}}^\top \right) \left(\prod_{j=0}^{n-1} \sum_{u_0[j], \dots, u_{2^d-1}[j] \in \mathbb{F}_2} \sigma[\tau_j(\varpi_j(\mathbf{u}))] \bigotimes_{k=0}^{2^d-1} M_{\lambda_0^{(k)}[j], \lambda_1^{(k)}[j]} \right) \left(\bigotimes_{j=0}^{2^d-1} \mathbf{e}_{i^{(j)}} \right)
 \end{aligned}$$

HDL Application Results

Applications to second-order DL distinguishers

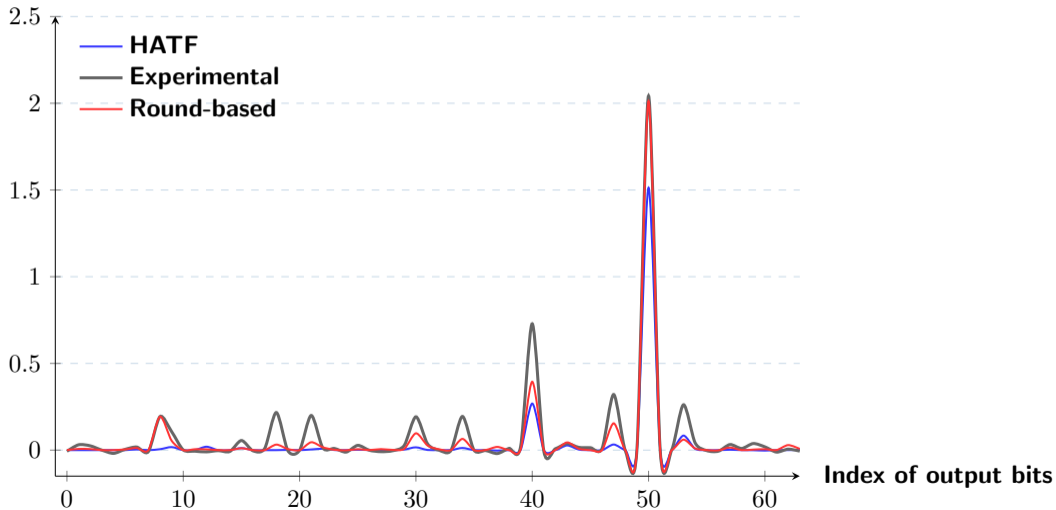
Target	Rounds	Exp. Cor.	Th. Cor.	Method	Ref.
Second-order differential-linear distinguisher					
Ascon-128 init.	5	$2^{-5.60}$	$2^{-6.05}$ $2^{-5.63}$	HATF Round-based	[HPT+23] [Ours]
	6 ★		$2^{-45.35}$	Round-based	[Ours]
Subterranean-2.0	5 ★	$2^{-6.05}$	$2^{-6.05}$	Round-based	[Ours]
	7 ★		$2^{-90.99}$	Round-based	[Ours]
	8 ★ (full)		$2^{-63.84}$	Round-based	[Ours]
Koala-p	5 ★	$2^{-5.89}$	$2^{-6.09}$	Round-based	[Ours]
	7 ★		$2^{-86.24}$	Round-based	[Ours]
	8 ★ (full)		$2^{-63.75}$	Round-based	[Ours]

A possible explanation for counter-intuitive 7-round and 8-round correlations

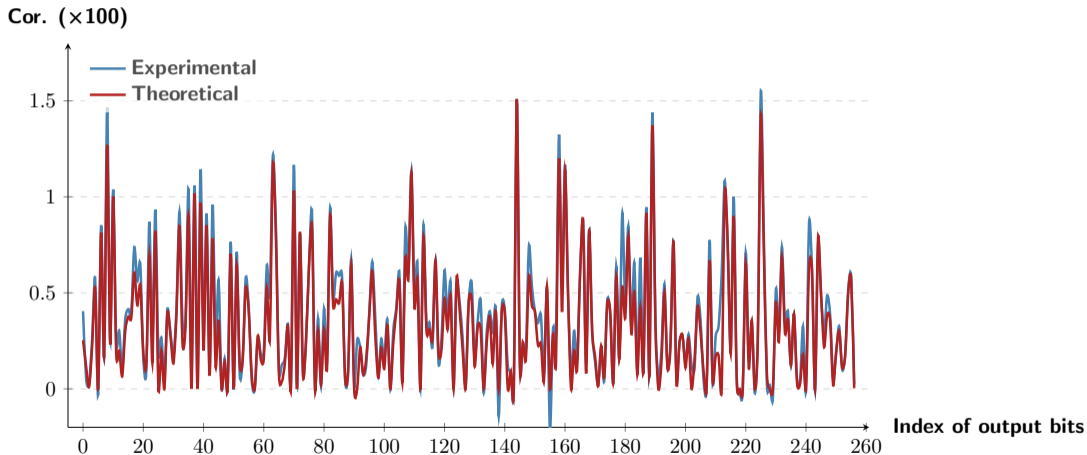
$$\gamma_r[(0, \underbrace{\lambda, \dots, \lambda}_{2^d-1})] \text{ is NOT ONLY from } \gamma_{r-1}[(0, \underbrace{\lambda, \dots, \lambda}_{2^d-1})] \text{ when } d > 1$$

Precision of Second-Order HDL Correlation for 5-Round Ascon

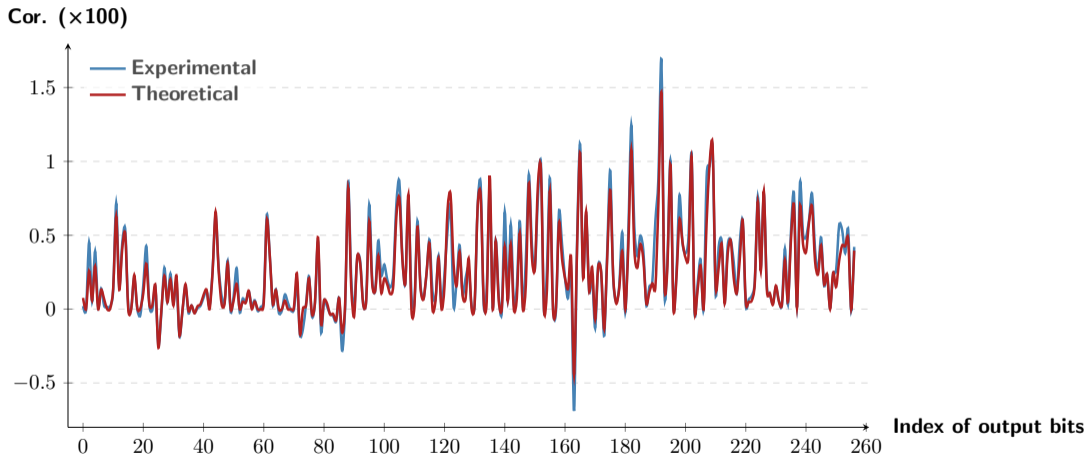
Cor. ($\times 100$)



Precision of Second-Order HDL Correlation for 5-Round Subterranean-2.0



Precision of Second-Order HDL Correlation for 5-Round Koala-p



Classification of DL Approximation Methods

Classification



- **Quasidifferential** basis for input, **2-wise linear** basis for output

Task: to **approximate** $Q^F[(0, \lambda), (0, \Delta)]$

- According to change-of-basis timing:
 - At the **beginning**: the current method [**Ours**], ATF [**LLL21**]
 - In the **middle**: classical 2-phase [**LH94**], 3-phase method, DLCT [**BDK+19**], generalized DLCT [**HDE+24**]
 - At the **last**: truncated differential table [**PZW+24**]
- According to the approximation method
 - **Round**-based: ATF [**LLL21**], the current method [**Ours**], TDT [**PZW+24**]

$$Q^F[(0, \lambda), (0, \Delta)] = \lambda^T \cdot M_{r-1} \cdot M_{r-2} \cdots M_0 \cdot \sigma_0$$

- **Trail**-based: classical 2-phase [**LH94**], 3-phase method, DLCT [**BDK+19**], generalized DLCT [**HDE+24**]

$$Q^F[(0, \lambda), (0, \Delta)] = \sum_{(0, \Delta) = (v_0, v_1, v_2, \dots, v_{r-1}, v_r = (0, \lambda))} \prod_{j=0}^{r-1} \omega_{v_{j+1}, v_j}$$

Classification



- **Quasidifferential** basis for input, **2-wise linear** basis for output

Task: to **approximate** $Q^F[(0, \lambda), (0, \Delta)]$

- According to change-of-basis timing:
 - At the **beginning**: the current method [Ours], ATF [LLL21]
 - In the **middle**: classical 2-phase [LH94], 3-phase method, DLCT [BDK+19], generalized DLCT [HDE+24]
 - At the **last**: truncated differential table [PZW+24]
- According to the approximation method
 - **Round**-based: ATF [LLL21], the current method [Ours], TDT [PZW+24]

$$Q^F[(0, \lambda), (0, \Delta)] = \lambda^T \cdot M_{r-1} \cdot M_{r-2} \cdots M_0 \cdot \sigma_0$$

- **Trail**-based: classical 2-phase [LH94], 3-phase method, DLCT [BDK+19], generalized DLCT [HDE+24]

$$Q^F[(0, \lambda), (0, \Delta)] = \sum_{(0, \Delta) = v_0, v_1, v_2, \dots, v_{r-1}, v_r = (0, \lambda)} \prod_{j=0}^{r-1} \omega_{v_{j+1}, v_j}$$

Thank you for your attention!

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