

Linear and differential cryptanalysis

Tim Beyne

`tim@cryptanalysis.info`

KU Leuven


March 11, 2025

The logo for KU Leuven, consisting of a dark blue rectangle with the text "KU LEUVEN" in white, bold, uppercase letters.

KU LEUVEN

Differential cryptanalysis

Overview

- ▶ Differentials and differential characteristics
- ▶ Quasidifferential transition matrices
- ▶ Quasidifferential trails
- ▶ Cost analysis
- ▶ Key-recovery techniques
-  Exercises

Differentials

- ▶ **Probabilistic** relation between an input difference a and an output difference b

$$F(x + a) \approx F(x) + b$$

- ▶ Pair (a, b) of differences $a \in \mathbb{F}_2^n$ and $b \in \mathbb{F}_2^m$ determines the differential

Differentials

- ▶ **Probabilistic** relation between an input difference a and an output difference b

$$F(x + a) \approx F(x) + b$$

- ▶ Pair (a, b) of differences $a \in \mathbb{F}_2^n$ and $b \in \mathbb{F}_2^m$ determines the differential
- ▶ If F is a uniform random function, then the number of inputs x such that $F(x + a) = F(x) + b$ is $2^n/2^m$ on average
- ▶ Probability of a differential:

$$p = \frac{\#\{x \in \mathbb{F}_2^n \mid F(x + a) = F(x) + b\}}{2^n} = \Pr_{\mathbf{x}} [F(\mathbf{x} + a) = F(\mathbf{x}) + b]$$

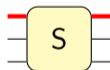
Differentials

Example

- ▶ 3-bit S-box $S: \mathbb{F}_2^3 \rightarrow \mathbb{F}_2^3$

| | | | | | | | | |
|--------|-----|-----|-----|-----|-----|-----|-----|-----|
| x | 000 | 001 | 010 | 011 | 100 | 101 | 110 | 111 |
| $S(x)$ | 111 | 010 | 100 | 101 | 001 | 110 | 011 | 000 |

- ▶ Differential $(a, b) = (001, 001)$



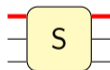
Differentials

Example

- ▶ 3-bit S-box $S: \mathbb{F}_2^3 \rightarrow \mathbb{F}_2^3$

| | | | | | | | | |
|--------|-----|-----|-----|-----|-----|-----|-----|-----|
| x | 000 | 001 | 010 | 011 | 100 | 101 | 110 | 111 |
| $S(x)$ | 111 | 010 | 100 | 101 | 001 | 110 | 011 | 000 |

- ▶ Differential $(a, b) = (001, 001)$



- ▶ Probability $\Pr_{\mathbf{x}} [S(\mathbf{x} + a) = S(\mathbf{x}) + b] = \frac{2}{8} = \frac{1}{4}$

Differentials

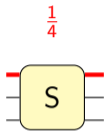
Distinguishers

- ▶ Sample q input pairs $(x_1, x_1 + a), \dots, (x_q, x_q + a)$ at random
- ▶ Average number of pairs with output difference b is pq
- ▶ $q \approx 1/p$ samples are enough for a distinguisher because right pairs are uncommon (assuming p is not too small or large)

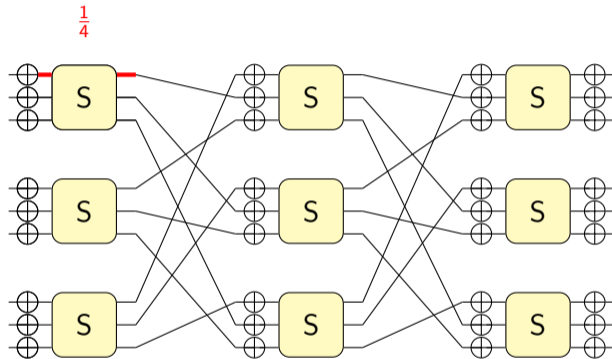


Number of samples depends on true- and false-positive probabilities (see later)

Differentials



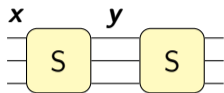
Differentials



Propagation through a sequence of operations?

Differentials

Example



$$\Pr[S^2(\mathbf{x} + a) = S^2(\mathbf{x}) + b] \approx \Pr[S(\mathbf{x} + a) = S(\mathbf{x}) + c \text{ and } S(\mathbf{y} + c) = S(\mathbf{y}) + b]$$

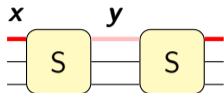


Pretend that \mathbf{x} and \mathbf{y} are independent:

$$\Pr[S^2(\mathbf{x} + a) = S^2(\mathbf{x}) + b] \overset{\text{skull}}{\approx} \Pr[S(\mathbf{x} + a) = S(\mathbf{x}) + c] \times \Pr[S(\mathbf{y} + c) = S(\mathbf{y}) + b]$$

Differentials

Example



$$\Pr[S^2(\mathbf{x} + a) = S^2(\mathbf{x}) + b] \approx \Pr[S(\mathbf{x} + a) = S(\mathbf{x}) + c \text{ and } S(\mathbf{y} + c) = S(\mathbf{y}) + b]$$



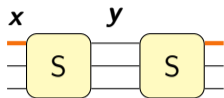
Pretend that \mathbf{x} and \mathbf{y} are independent:

$$\Pr[S^2(\mathbf{x} + a) = S^2(\mathbf{x}) + b] \overset{\text{skull}}{\approx} \Pr[S(\mathbf{x} + a) = S(\mathbf{x}) + c] \times \Pr[S(\mathbf{y} + c) = S(\mathbf{y}) + b]$$

- ▶ For example: $a = b = c = 001$ gives $1/4 \times 1/4 = 1/16$
- ▶ Unfortunately, this is wrong (the correct result is $1/4$)

Differentials

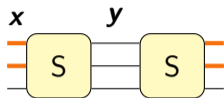
Example



$$\begin{aligned} & \Pr[S^2(\mathbf{x} + 001) = S^2(\mathbf{x}) + 001] \\ &= \Pr[S(\mathbf{x} + 001) = S(\mathbf{x}) + 001 \text{ and } S(\mathbf{y} + 001) = S(\mathbf{y}) + 001] + \\ & \Pr[S(\mathbf{x} + 001) = S(\mathbf{x}) + 011 \text{ and } S(\mathbf{y} + 011) = S(\mathbf{y}) + 001] + \\ & \Pr[S(\mathbf{x} + 001) = S(\mathbf{x}) + 101 \text{ and } S(\mathbf{y} + 101) = S(\mathbf{y}) + 001] + \\ & \Pr[S(\mathbf{x} + 001) = S(\mathbf{x}) + 111 \text{ and } S(\mathbf{y} + 111) = S(\mathbf{y}) + 001] \\ & \approx \frac{1}{16} + \frac{1}{16} + \frac{1}{16} + \frac{1}{16} \end{aligned}$$

Differentials

Example



$$\begin{aligned} & \Pr[S^2(\mathbf{x} + 011) = S^2(\mathbf{x}) + 011] \\ &= \Pr[S(\mathbf{x} + 011) = S(\mathbf{x}) + 001 \text{ and } S(\mathbf{y} + 001) = S(\mathbf{y}) + 011] + \\ & \Pr[S(\mathbf{x} + 011) = S(\mathbf{x}) + 010 \text{ and } S(\mathbf{y} + 010) = S(\mathbf{y}) + 011] + \\ & \Pr[S(\mathbf{x} + 011) = S(\mathbf{x}) + 101 \text{ and } S(\mathbf{y} + 101) = S(\mathbf{y}) + 011] + \\ & \Pr[S(\mathbf{x} + 001) = S(\mathbf{x}) + 110 \text{ and } S(\mathbf{y} + 110) = S(\mathbf{y}) + 001] \\ & \approx \frac{1}{16} + \frac{1}{16} + \frac{1}{16} + \frac{1}{16} \end{aligned}$$

- ▶ Unfortunately, this is still wrong (the correct result is 0)
- ▶ It is not reasonable to assume independence

Differential characteristics

- ▶ Suppose $F = F_r \circ \dots \circ F_2 \circ F_1$ and let $\mathbf{x}_i = F_i(\mathbf{x}_{i-1})$ with $\mathbf{x}_0 = \mathbf{x}$
- ▶ Law of total probability:

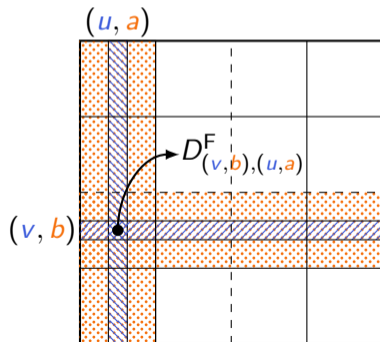
$$\Pr[F(\mathbf{x} + \mathbf{a}_1) = F(\mathbf{x}) + \mathbf{a}_{r+1}] = \sum_{\mathbf{a}_2, \dots, \mathbf{a}_r} \Pr[\bigwedge_{i=1}^r F_i(\mathbf{x}_i + \mathbf{a}_i) = F(\mathbf{x}_i) + \mathbf{a}_{i+1}]$$

- ▶ A sequence $(\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_{r+1})$ is called a differential characteristic
- ▶ How to calculate the probability of a characteristic?

Quasidifferential transition matrices

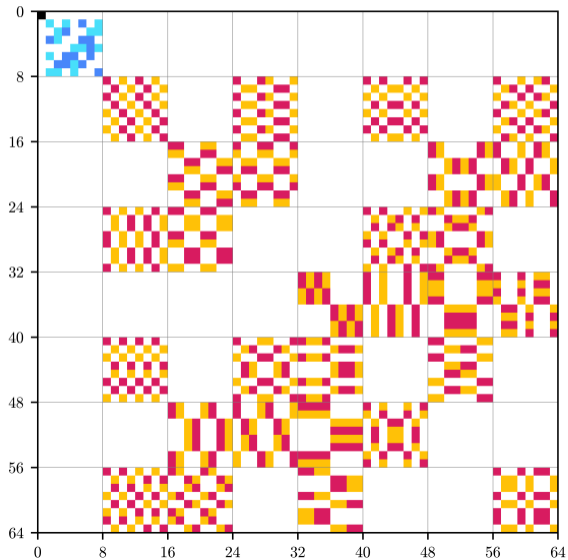
- ▶ $2^{2m} \times 2^{2n}$ matrix corresponding to $F: \mathbb{F}_2^n \rightarrow \mathbb{F}_2^m$

$$D_{(v,b),(u,a)}^F = \left(2 \Pr_x [v^T F(x) = u^T x \mid F(x+a) = F(x) + b] - 1 \right) \\ \times \Pr_x [F(x+a) = F(x) + b]$$



Quasidifferential transition matrices

Example



Quasidifferential transition matrices

Multiplication property

- ▶ If $F = F_2 \circ F_1$, then

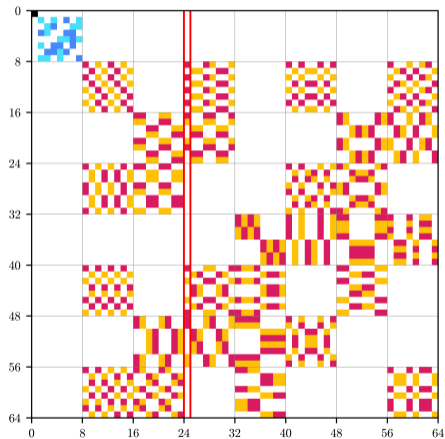
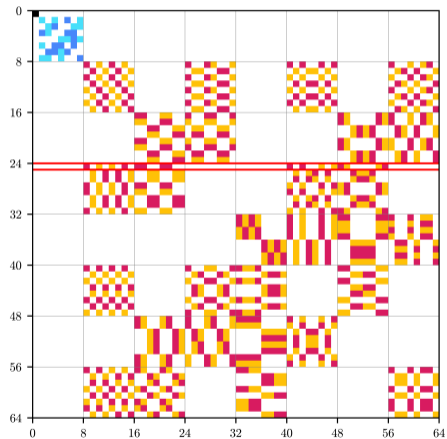
$$D^F = D^{F_2} D^{F_1}$$

 Proof by calculation

- ▶ This is the most important property of quasidifferential transition matrices
- ▶ There are more conceptual (but more abstract) proofs without calculation

Quasidifferential transition matrices

Multiplication property: example



Quasidifferential trails

- ▶ If $F = F_r \circ \dots \circ F_2 \circ F_1$, then $D^F = D^{F_r} \dots D^{F_2} D^{F_1}$, so

$$D_{\varpi_{r+1}, \varpi_1}^F = \sum_{\varpi_2, \dots, \varpi_r} D_{\varpi_{r+1}, \varpi_r}^{F_r} \dots D_{\varpi_3, \varpi_2}^{F_2} D_{\varpi_2, \varpi_1}^{F_1}$$

with $\varpi_i = (u_i, a_i)$ for $i \in \{1, \dots, r\}$

- ▶ A quasidifferential trail is a sequence $(\varpi_1, \dots, \varpi_{r+1})$ with correlation

$$\prod_{i=1}^r D_{\varpi_{i+1}, \varpi_i}^{F_i}$$

- ▶ Analysis relies on the assumption that there exists a set Λ of 'dominant trails':

$$D_{\varpi_{r+1}, \varpi_1}^F = \sum_{\varpi \in \Lambda} \prod_{i=1}^r D_{\varpi_{i+1}, \varpi_i}^{F_i} + \varepsilon$$

Quasidifferential trails

- ▶ $D_{(0,a_{r+1}), (0,a_1)}^F$ is the probability of the differential (a_1, a_{r+1})
- ▶ Quasidifferential trails can be used to compute the probability of a differential
- ▶ Quasidifferential trails can be used to compute the probability of a characteristic:

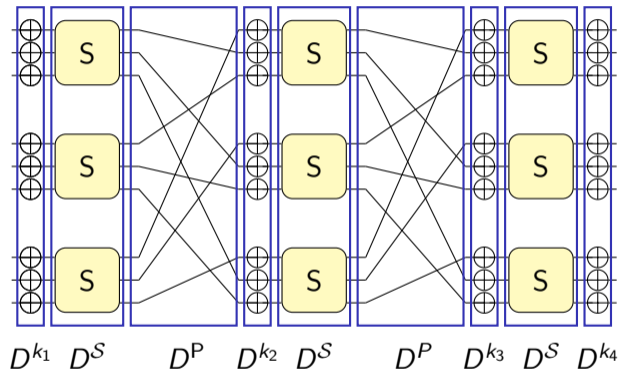
$$\sum_{u_2, \dots, u_r} \prod_{i=1}^r D_{(u_{i+1}, a_{i+1}), (u_i, a_i)}^{F_i}$$

Proof: similar as for the multiplication property (exercise)

visual proof ()

Quasidifferential trails

Example



- ▶ To analyze trails we need to determine D^{k_i} , D^S and D^P

Quasidifferential trails

Bricklayer functions

- ▶ If $F(x_1 \| x_2) = F_1(x_1) \| F_2(x_2)$, then

$$D_{(v_1 \| v_2, b_1 \| b_2), (u_1 \| u_2, a_1 \| a_2)}^F = D_{(v_1, b_1), (u_1, a_1)}^{F_1} D_{(v_2, b_2), (u_2, a_2)}^{F_2}$$

 Proof by calculation

- ▶ Equivalently, $D^F = D^{F_1} \otimes D^{F_2}$
- ▶ For the S-box layer: $D^S = D^S \otimes D^S \otimes D^S$

Quasidifferential trails

Translations and linear functions

- ▶ If $F(x) = x + k$, then

$$D_{(v,b),(u,a)}^F = \begin{cases} (-1)^{v^T k} & \text{if } u = v \text{ and } a = b \\ 0 & \text{else.} \end{cases}$$

 Proof

Quasidifferential trails

Translations and linear functions

- ▶ If $F(x) = x + k$, then

$$D_{(v,b),(u,a)}^F = \begin{cases} (-1)^{v^T k} & \text{if } u = v \text{ and } a = b \\ 0 & \text{else.} \end{cases}$$

 Proof

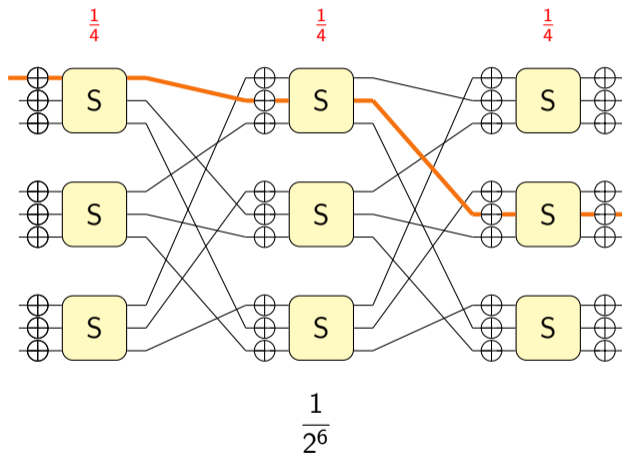
- ▶ If $F(x) = Mx$, then

$$D_{(v,b),(u,a)}^F = \begin{cases} 1 & \text{if } u = M^T v \text{ and } b = Ma \\ 0 & \text{else.} \end{cases}$$

 Proof

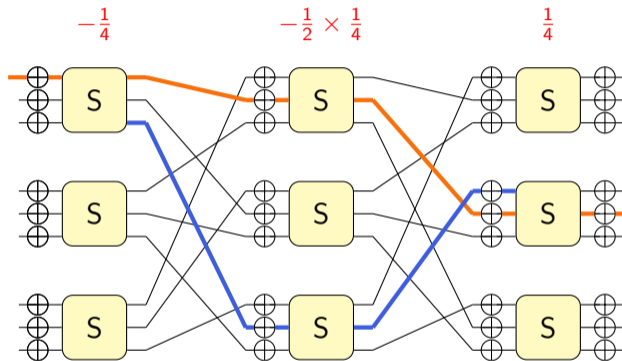
Quasidifferential trails

Example: 3-round differential (characteristic 1)



Quasidifferential trails

Example: 3-round differential (characteristic 1)

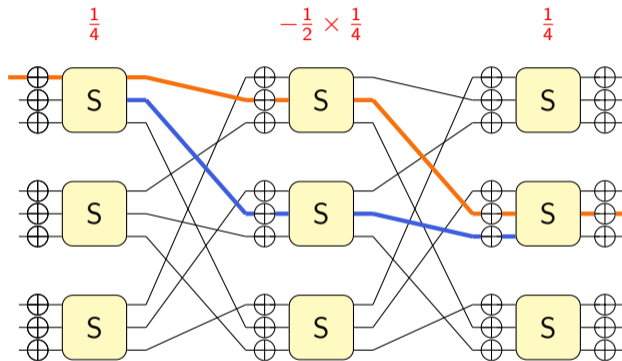


$$\frac{1}{2^6} + (-1)^{\kappa_1} \frac{1}{2^7}$$

$$\text{with } \kappa_1 = k_{2,8} + k_{3,4}$$

Quasidifferential trails

Example: 3-round differential (characteristic 1)

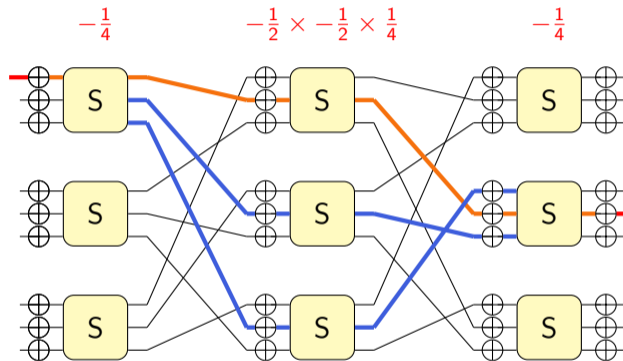


$$\frac{1}{2^6} + (-1)^{\kappa_1} \frac{1}{2^7} + (-1)^{\kappa_2} \frac{1}{2^7}$$

with $\kappa_1 = k_{2,8} + k_{3,4}$, $\kappa_2 = k_{2,5} + k_{3,6}$

Quasidifferential trails

Example: 3-round differential (characteristic 1)

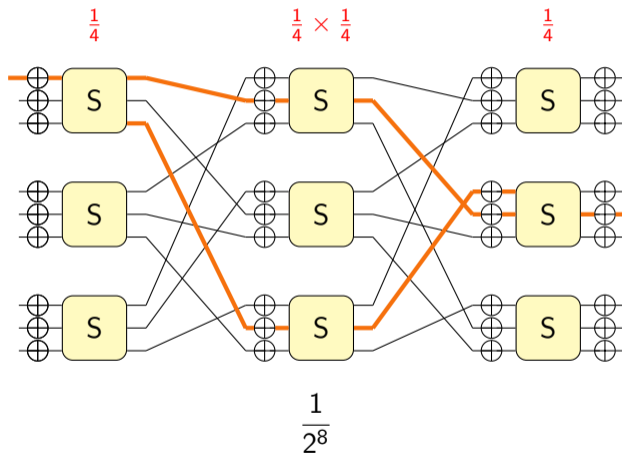


$$\frac{1}{2^6} + (-1)^{\kappa_1} \frac{1}{2^7} + (-1)^{\kappa_2} \frac{1}{2^7} + (-1)^{\kappa_1 + \kappa_2} \frac{1}{2^8}$$

$$\text{with } \kappa_1 = k_{2,8} + k_{3,4}, \kappa_2 = k_{2,5} + k_{3,6}$$

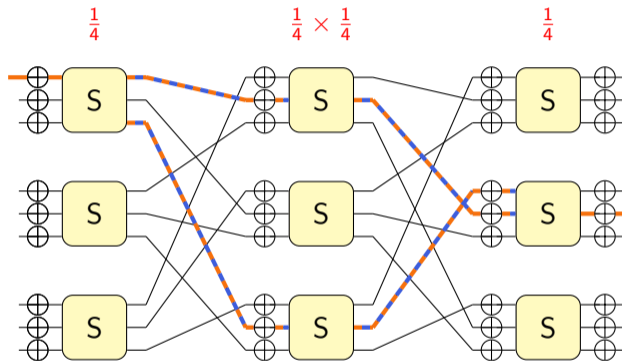
Quasidifferential trails

Example: 3-round differential (characteristic 2)



Quasidifferential trails

Example: 3-round differential (characteristic 2)

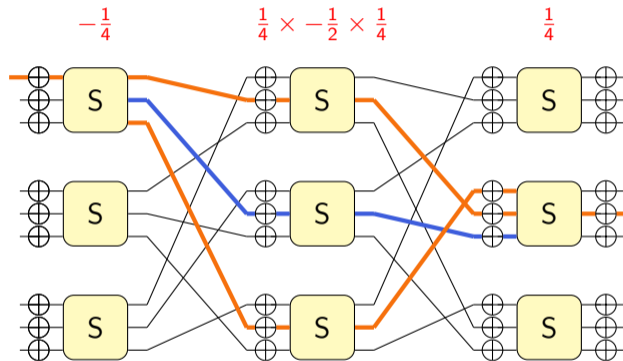


$$\frac{1}{2^8} + (-1)^{\kappa_1 + \kappa_3} \frac{1}{2^8}$$

with $\kappa_1 = k_{2,8} + k_{3,4}$, $\kappa_3 = k_{2,2} + k_{3,5}$

Quasidifferential trails

Example: 3-round differential (characteristic 2)

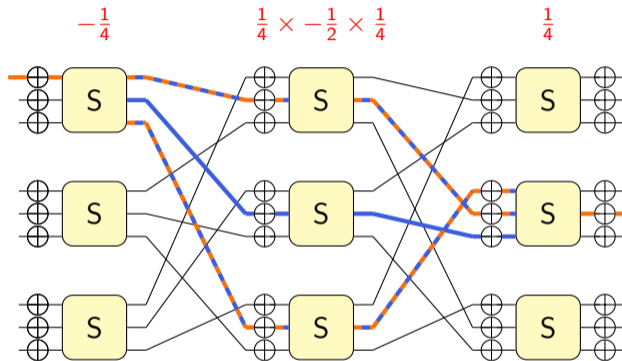


$$\frac{1}{2^8} + (-1)^{\kappa_1 + \kappa_3} \frac{1}{2^8} + (-1)^{\kappa_2} \frac{1}{2^9}$$

with $\kappa_1 = k_{2,8} + k_{3,4}$, $\kappa_2 = k_{2,5} + k_{3,6}$, $\kappa_3 = k_{2,2} + k_{3,5}$

Quasidifferential trails

Example: 3-round differential (characteristic 2)

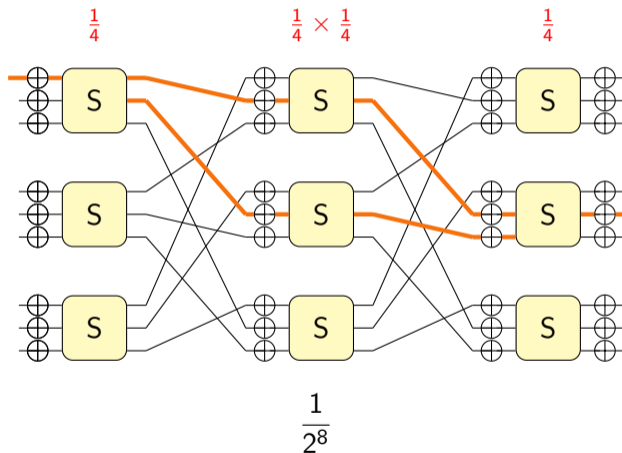


$$\frac{1}{2^8} + (-1)^{\kappa_1 + \kappa_3} \frac{1}{2^8} + (-1)^{\kappa_2} \frac{1}{2^9} + (-1)^{\kappa_1 + \kappa_2 + \kappa_3} \frac{1}{2^9}$$

with $\kappa_1 = k_{2,8} + k_{3,4}$, $\kappa_2 = k_{2,5} + k_{3,6}$, $\kappa_3 = k_{2,2} + k_{3,5}$

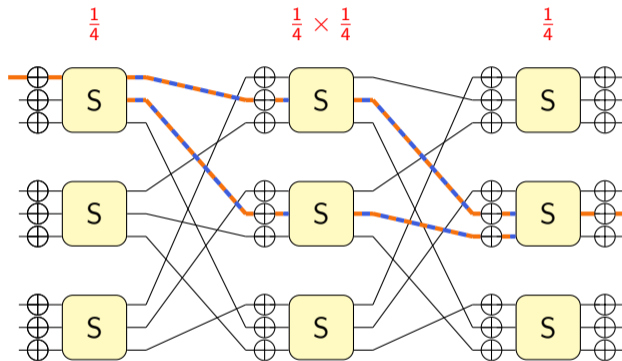
Quasidifferential trails

Example: 3-round differential (characteristic 3)



Quasidifferential trails

Example: 3-round differential (characteristic 3)

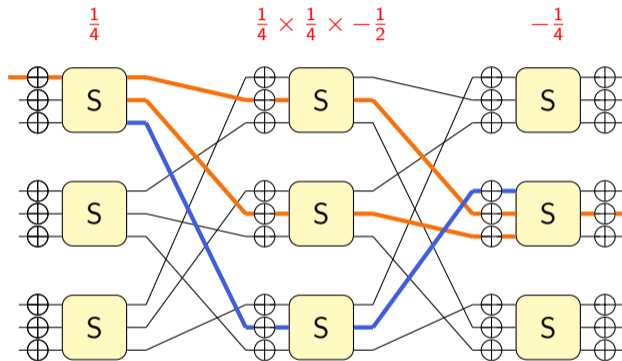


$$\frac{1}{2^8} + (-1)^{\kappa_2 + \kappa_3} \frac{1}{2^8}$$

with $\kappa_2 = k_{2,5} + k_{3,6}$, $\kappa_3 = k_{2,2} + k_{3,5}$

Quasidifferential trails

Example: 3-round differential (characteristic 3)

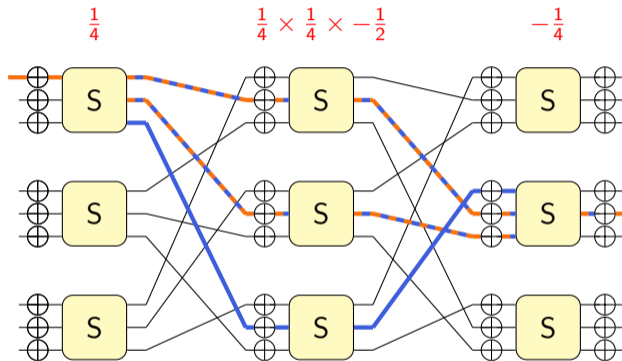


$$\frac{1}{2^8} + (-1)^{\kappa_2 + \kappa_3} \frac{1}{2^8} + (-1)^{\kappa_1} \frac{1}{2^9}$$

with $\kappa_1 = k_{2,8} + k_{3,4}$, $\kappa_2 = k_{2,5} + k_{3,6}$, $\kappa_3 = k_{2,2} + k_{3,5}$

Quasidifferential trails

Example: 3-round differential (characteristic 3)

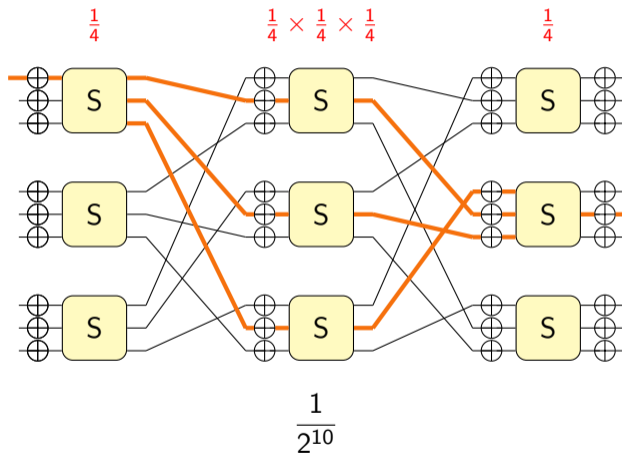


$$\frac{1}{2^8} + (-1)^{\kappa_2 + \kappa_3} \frac{1}{2^8} + (-1)^{\kappa_1} \frac{1}{2^9} + (-1)^{\kappa_1 + \kappa_2 + \kappa_3} \frac{1}{2^9}$$

with $\kappa_1 = k_{2,8} + k_{3,4}$, $\kappa_2 = k_{2,5} + k_{3,6}$, $\kappa_3 = k_{2,2} + k_{3,5}$

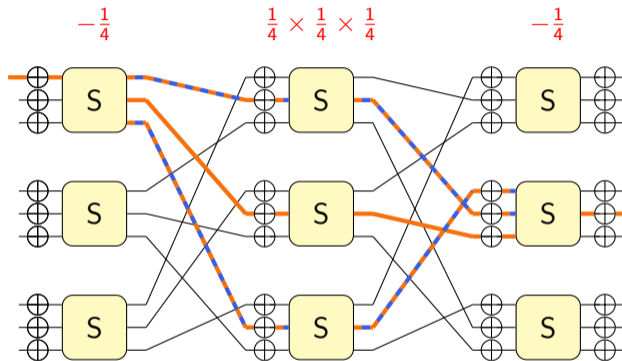
Quasidifferential trails

Example: 3-round differential (characteristic 4)



Quasidifferential trails

Example: 3-round differential (characteristic 4)

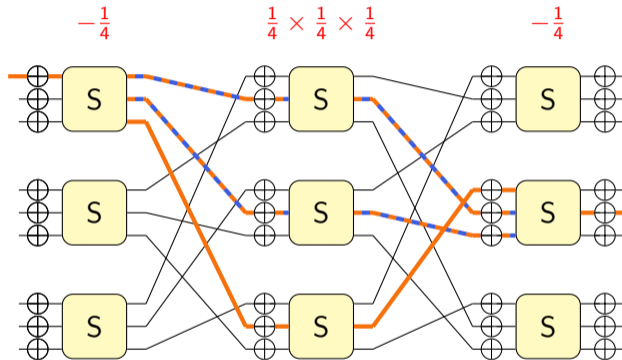


$$\frac{1}{2^{10}} + (-1)^{\kappa_1 + \kappa_3} \frac{1}{2^{10}}$$

with $\kappa_1 = k_{2,8} + k_{3,4}$, $\kappa_3 = k_{2,2} + k_{3,5}$

Quasidifferential trails

Example: 3-round differential (characteristic 4)

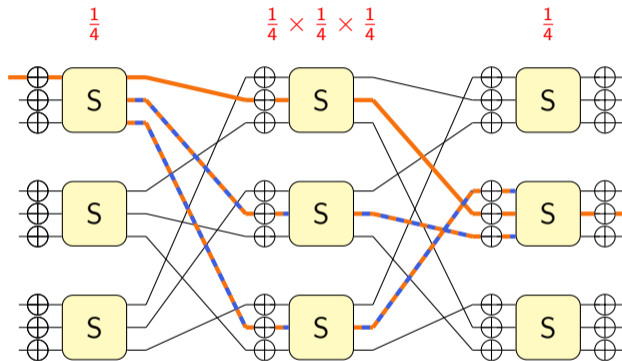


$$\frac{1}{2^{10}} + (-1)^{\kappa_1 + \kappa_3} \frac{1}{2^{10}} + (-1)^{\kappa_2 + \kappa_3} \frac{1}{2^{10}}$$

with $\kappa_1 = k_{2,8} + k_{3,4}$, $\kappa_2 = k_{2,5} + k_{3,6}$, $\kappa_3 = k_{2,2} + k_{3,5}$

Quasidifferential trails

Example: 3-round differential (characteristic 4)



$$\frac{1}{2^{10}} + (-1)^{\kappa_1 + \kappa_3} \frac{1}{2^{10}} + (-1)^{\kappa_2 + \kappa_3} \frac{1}{2^{10}} + (-1)^{\kappa_1 + \kappa_2} \frac{1}{2^{10}}$$

with $\kappa_1 = k_{2,8} + k_{3,4}$, $\kappa_2 = k_{2,5} + k_{3,6}$, $\kappa_3 = k_{2,2} + k_{3,5}$

Quasidifferential trails

Example: 3-round differential

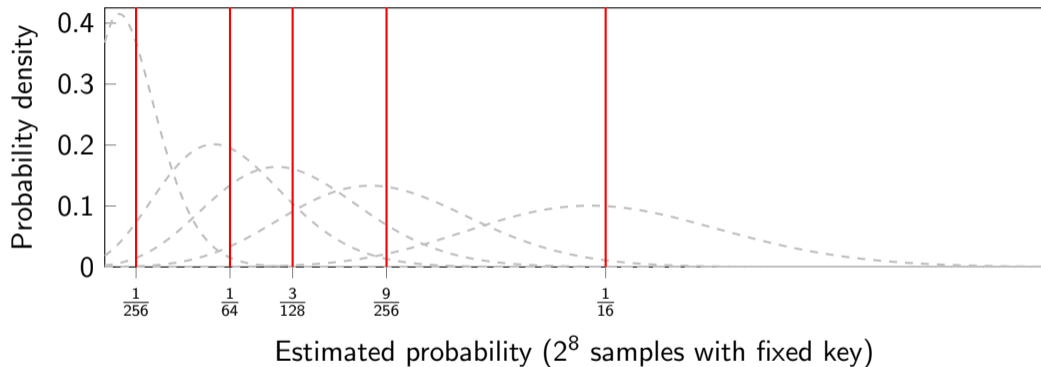
- ▶ Overall probability depends on three key bits

$$\begin{aligned} & \frac{1}{2^6} (1 + (-1)^{\kappa_1}/2)(1 + (-1)^{\kappa_2}/2) \\ & + \frac{1}{2^8} (1 + (-1)^{\kappa_1+\kappa_3})(1 + (-1)^{\kappa_1}/2) \\ & + \frac{1}{2^8} (1 + (-1)^{\kappa_2+\kappa_3})(1 + (-1)^{\kappa_2}/2) \\ & + \frac{1}{2^{10}} (1 + (-1)^{\kappa_1+\kappa_3})(1 + (-1)^{\kappa_2+\kappa_3}) \\ & \in \left\{ \frac{1}{256}, \frac{1}{64}, \frac{3}{128}, \frac{9}{256}, \frac{1}{16} \right\} \end{aligned}$$

- ⚠ Characteristics with ≥ 4 active S-boxes can contribute significantly

Quasidifferential trails

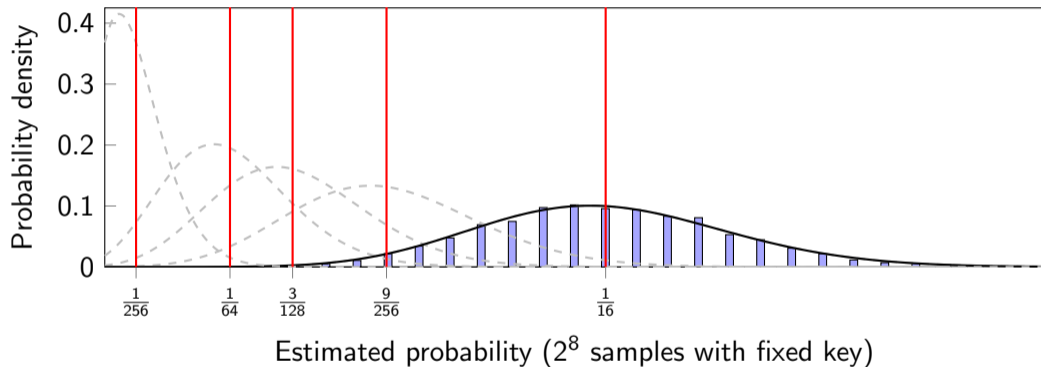
Example: 3-round differential



- Probability reveals something about the key (but we will see better methods later)

Quasidifferential trails

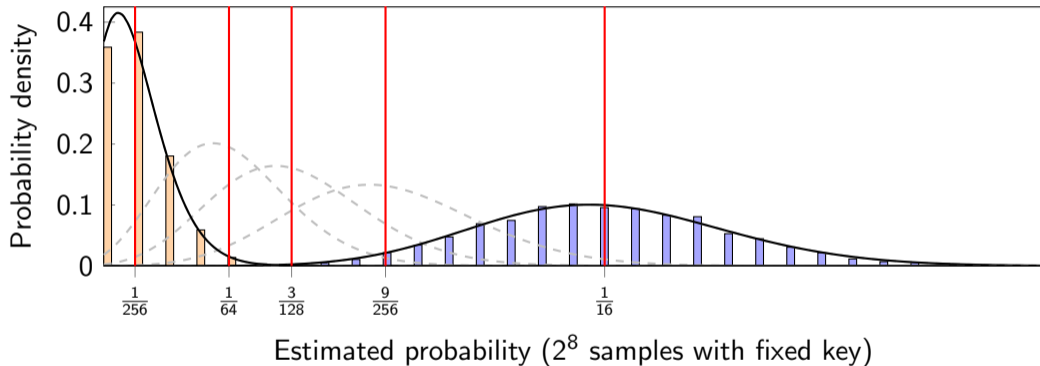
Example: 3-round differential



- Probability reveals something about the key (but we will see better methods later)

Quasidifferential trails

Example: 3-round differential



- Probability reveals something about the key (but we will see better methods later)

Cost analysis

- ▶ Using q independent samples $\mathbf{x}_1, \dots, \mathbf{x}_q$ (so $2q$ chosen plaintexts):

$$\hat{\mathbf{p}} = \frac{1}{q} \#\{1 \leq i \leq q \mid F(\mathbf{x}_i + a) = F(\mathbf{x}_i) + b\}$$

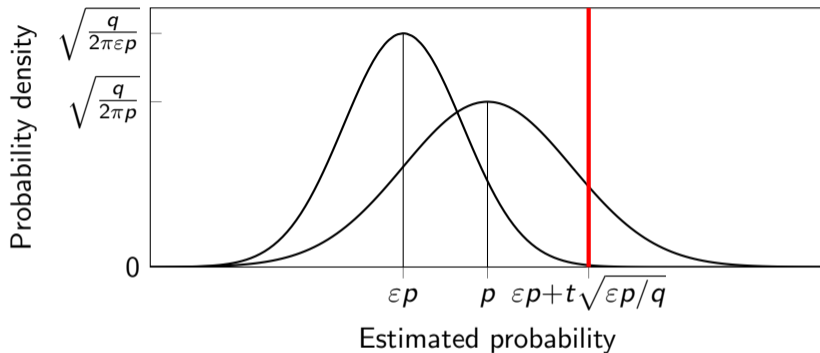
- ▶ Simplifications:

- pq is not too small and probability p is not too large
- Probability is εp for wrong keys

- ▶ Distribution of $\hat{\mathbf{p}}$ is close to normal with mean p and variance $p(1-p)/q \approx p/q$

- ▶ Hypothesis test: $\hat{\mathbf{p}} \geq \varepsilon p + t\sqrt{\varepsilon p/q}$

Cost analysis



True-positive probability: $P_S = \Phi((1 - \epsilon)\sqrt{pq} - t\sqrt{\epsilon})$

► False-positive probability: $P_F = \Phi(-t)$

Cost analysis

- ▶ Eliminating t gives

$$P_S = \Phi(\Phi^{-1}(P_F)\sqrt{\varepsilon} + (1 - \varepsilon)\sqrt{pq})$$

- ▶ Inverting this gives

$$q = \frac{1}{p} \left(\frac{\Phi^{-1}(P_S) - \Phi^{-1}(P_F)\sqrt{\varepsilon}}{1 - \varepsilon} \right)^2$$

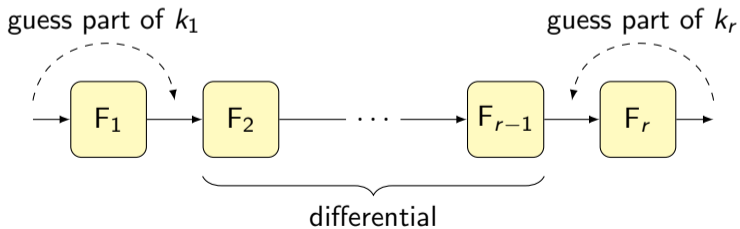
- ▶ If p depends on the key, need to average the formulas above
- ▶ $1/\varepsilon$ is sometimes called the 'signal-to-noise ratio'



This is essentially optimal *but important assumptions are made*

Key recovery

- ▶ If one characteristic is dominant:
 - (a) Differential probability depends on the key
 - (b) Part of the key can be deduced from the output difference
- ▶ Guessing key material from the first or last round is often more powerful



Key recovery

- ▶ Basic procedure
 - Count the number of right pairs per candidate key
 - Filter out invalid candidate keys using the hypothesis test
- ▶ For K candidate keys, $P_{\text{F}}K$ incorrect candidates remain
- ▶ Optimizations of the counting phase

Further topics

- ▶ Optimization of differential characteristics and quasidifferential trails
- ▶ Key-recovery techniques
- ▶ Multiple differentials
- ▶ Impossible differentials
- ▶ Truncated differentials
- ▶ Hash function cryptanalysis
- ▶ Geometric approach