

Summary

- ◆ Our proposed scheme can identify robustly images under various coding conditions.
- ◆ Quantization matrices and positions in which DCT coefficients have zero values are used as features.
- ◆ The features do not provide no visible information.
- ◆ The property of DCT coefficients and the features allow us to provide no false negative matches.
- ◆ Simulation results demonstrate the effectiveness of the proposed scheme.

Background

◆ What is "image identification"?

⇒ Identification of images which are generated from **the same original image** under various coding conditions.

◆ Why is image identification required?

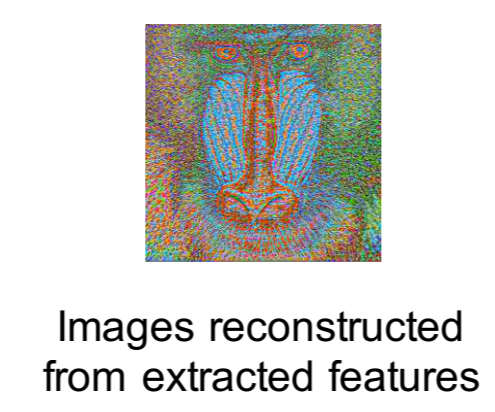
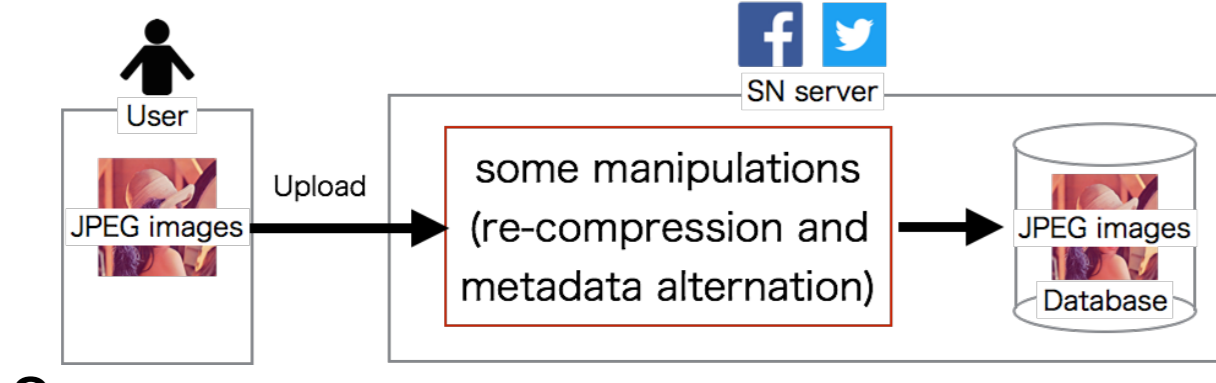
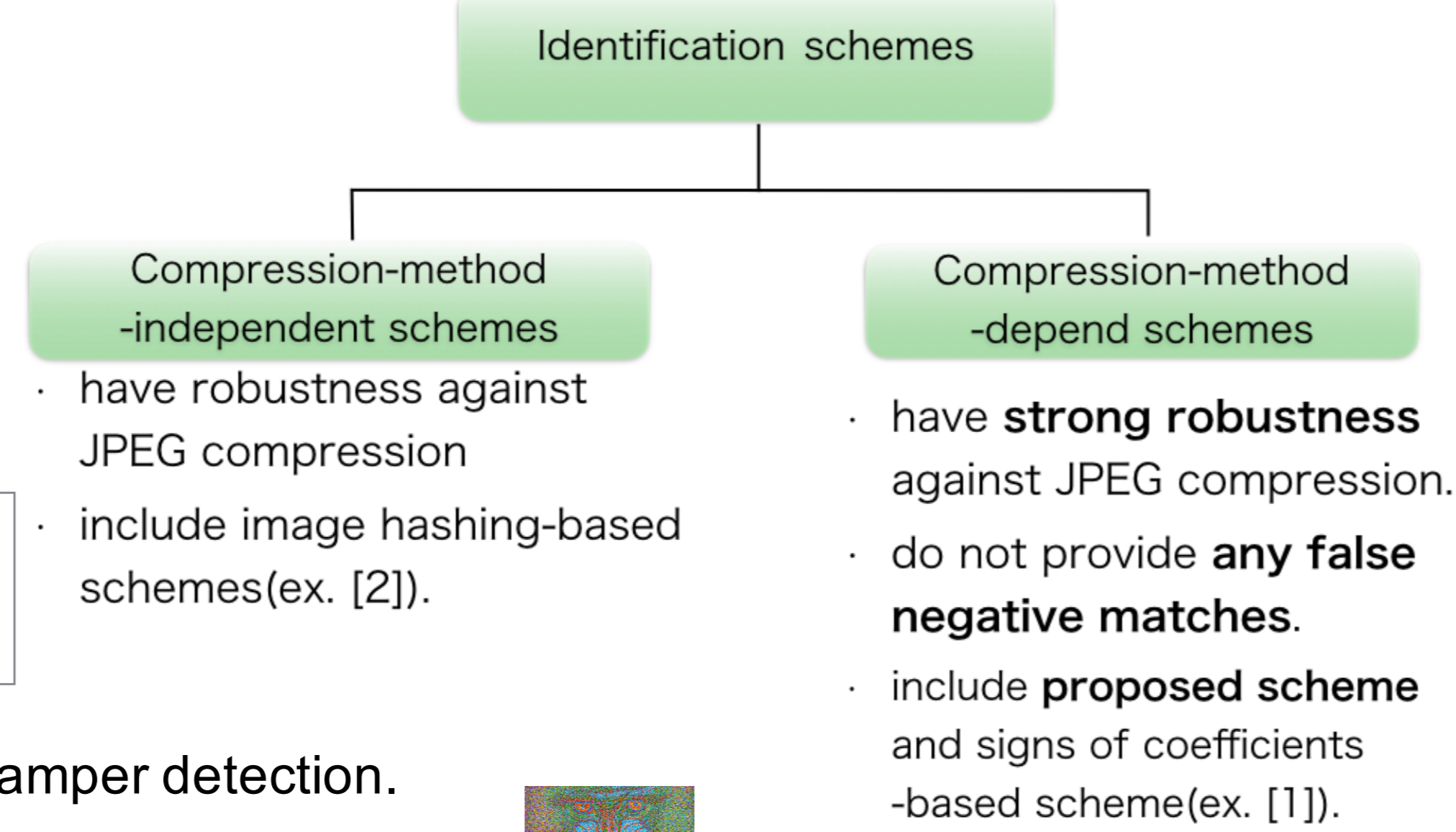
⇒ Social networks(SNs) providers often **re-compress** uploaded JPEG images with **the different coding parameter** from that of uploaded one.

◆ What are target applications?

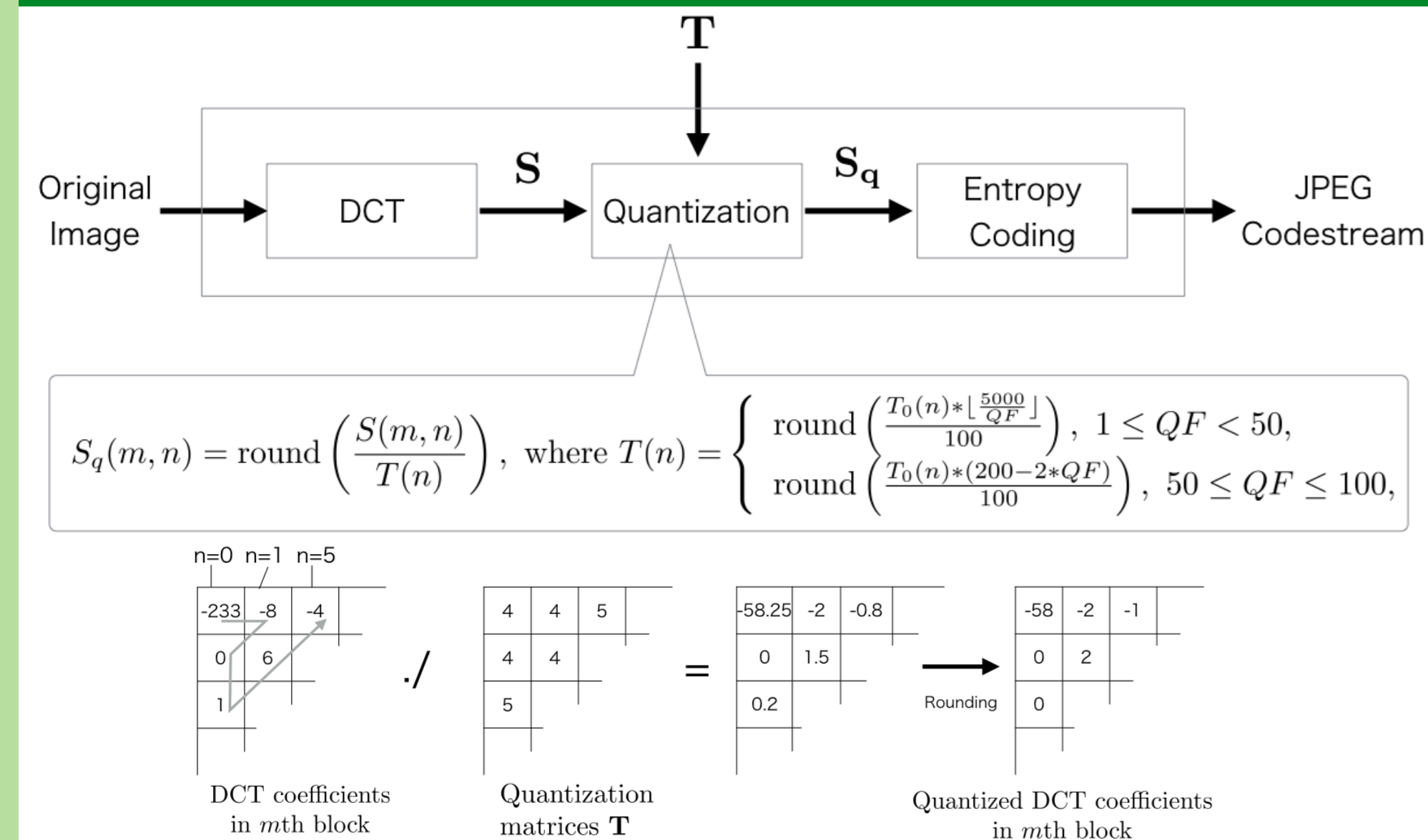
⇒ Relating images uploaded to SNs with downloaded ones and tamper detection.

◆ What limitations do conventional schemes using signs of DCT coefficients have?

- The features have to be protected because they have **visible information**.
- **Coding conditions** are limited.

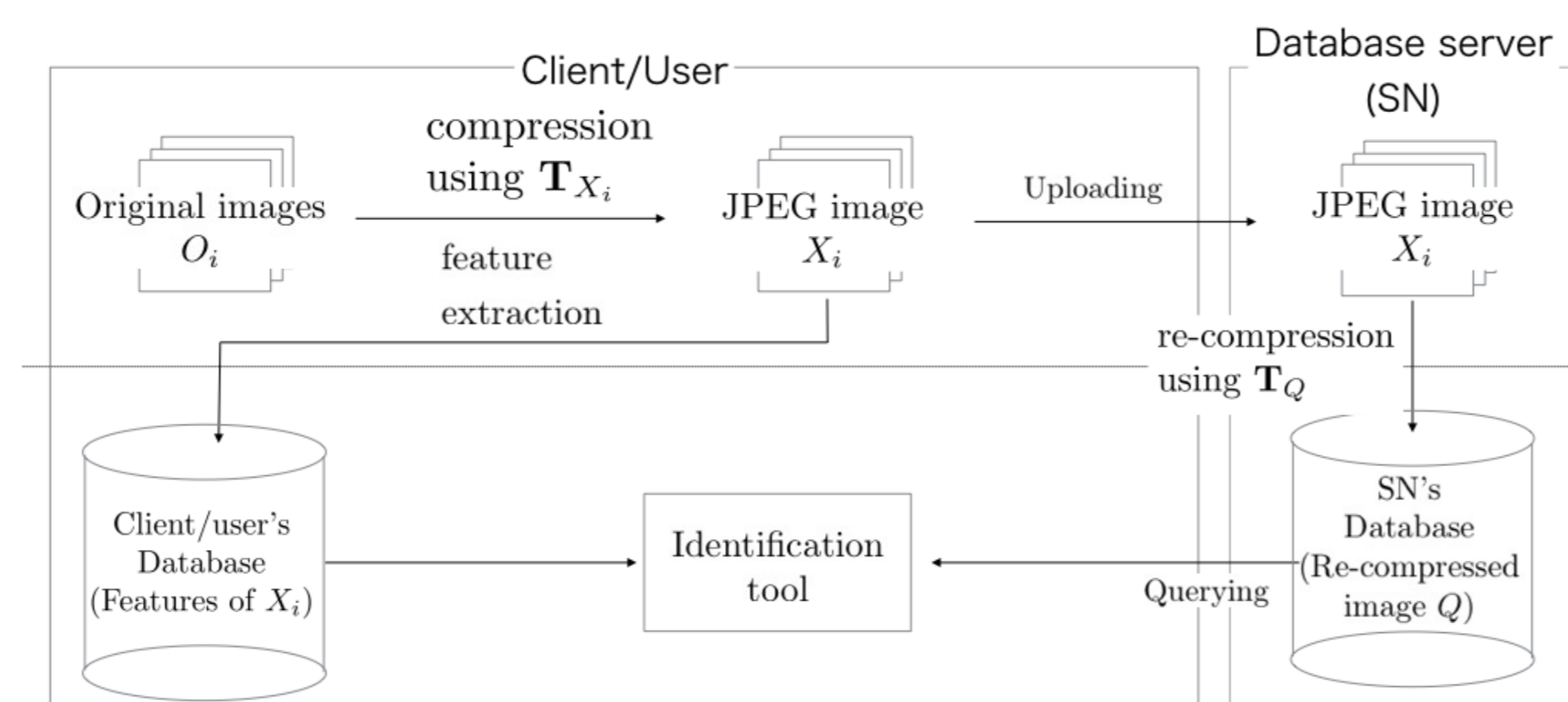


JPEG Compression

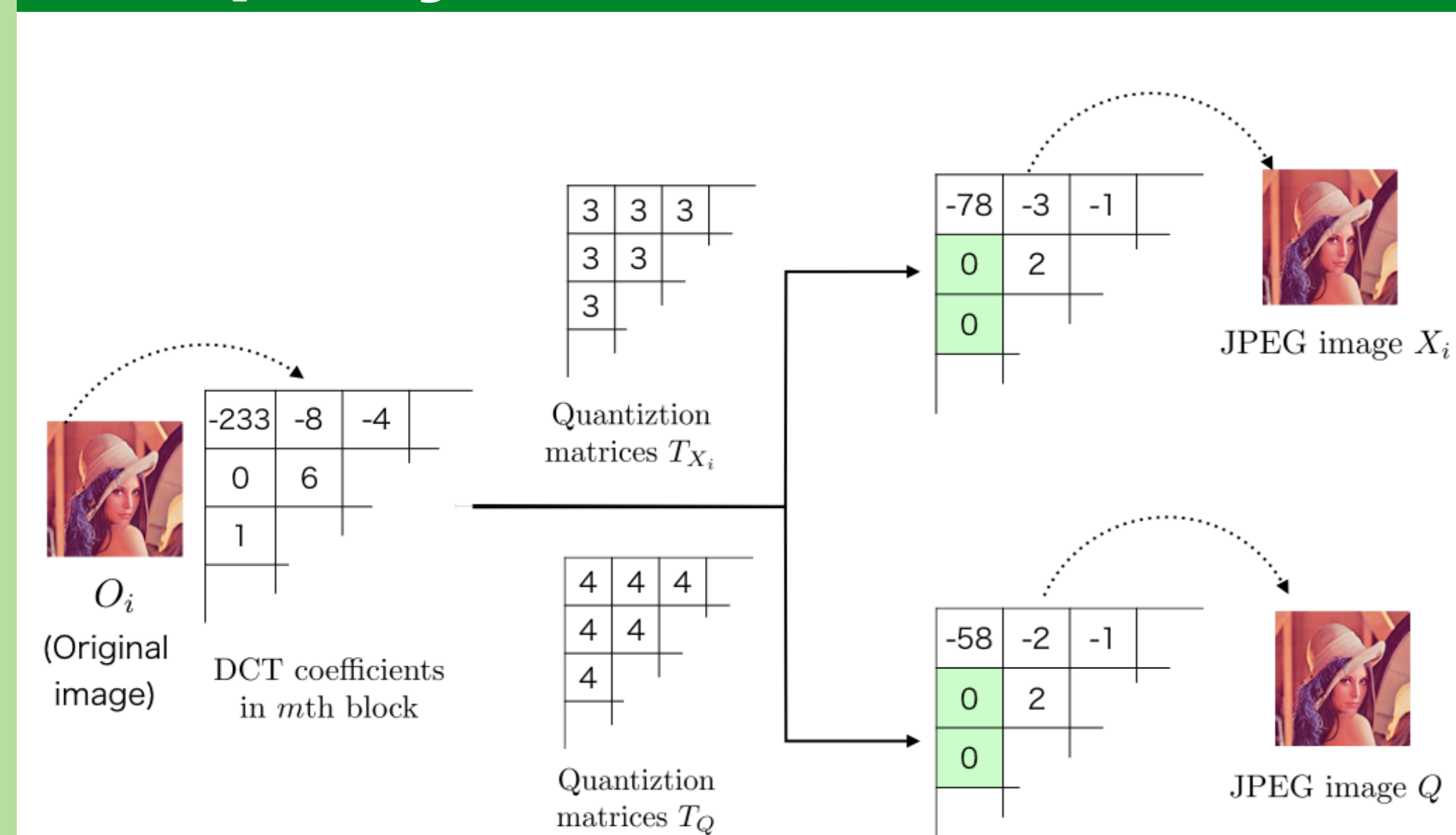


Scenario

- ◆ Images uploaded to SNs and downloaded ones are identified.
- ◆ Features have to provide **no visible information** due to privacy concerns or copyright protection.



Property of DCT coefficients



- ◆ When X_i and Q are generated from the same original image O_i ,

$$\begin{cases} Q(m, n) = 0, & \text{for } T_{X_i}(n) \leq T_Q(n) \text{ and } X_i(m, n) = 0, \\ X_i(m, n) = 0, & \text{for } T_{X_i}(n) \geq T_Q(n) \text{ and } Q(m, n) = 0, \end{cases}$$

where $\forall m \in \{0, \dots, M\}$ and $\forall n \in \{0, \dots, N\}$.

- ◆ X_i and Q are generated from the different original images, when

$$\begin{cases} Q(m, n) \neq 0, & \text{for } T_{X_i}(n) \leq T_Q(n) \text{ and } X_i(m, n) = 0, \\ X_i(m, n) \neq 0, & \text{for } T_{X_i}(n) \geq T_Q(n) \text{ and } Q(m, n) = 0, \end{cases} \quad (1)$$

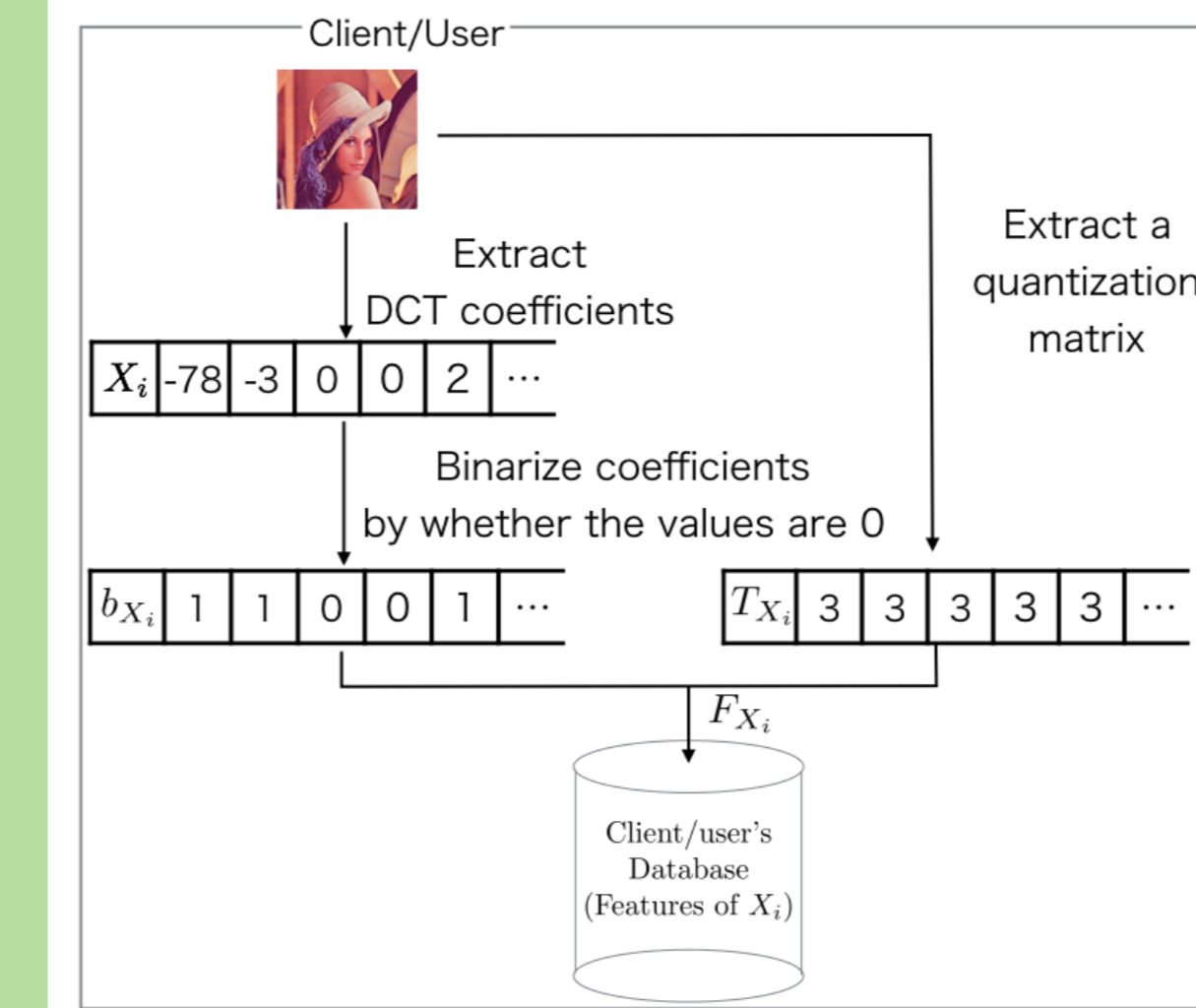
where $\exists m \in \{0, \dots, M\}$ and $\exists n \in \{0, \dots, N\}$.

- ◆ Proposed scheme uses Eq.(1) for identification.

Proposed Scheme

◆ Feature extraction process

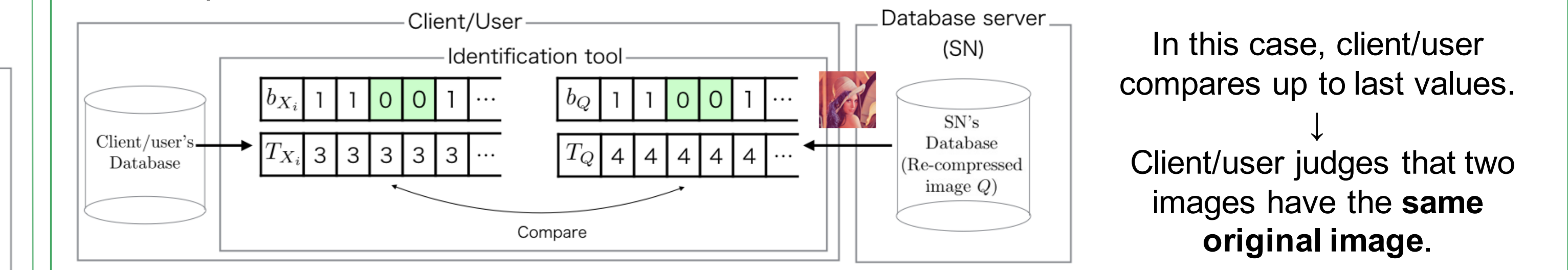
- Quantization matrices and positions of zero values are used as features.
- The set of features F_{X_i} provides no visible information.



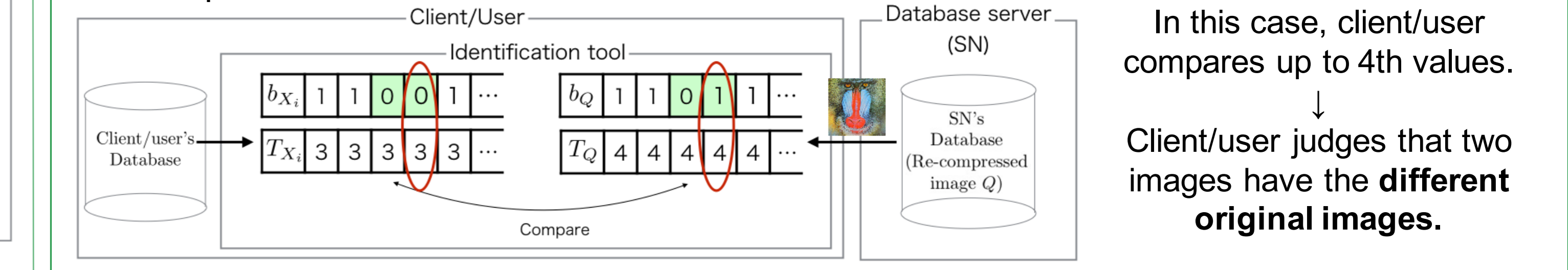
◆ Identification process

- Eq.(1) and features are used.
- Client/user confirms whether Eq.(1) is satisfied at each position.
 - not satisfied at all positions ⇒ Two images have **same original image**.
 - satisfied at a position ⇒ Two images have **different original images**.

Example 1



Example 2



Simulation

◆ Simulation condition

- Dataset :Head Posed Image Database (186 images in "Person01")
- Encoder: Independent JPEG Group(IJG)
- Identification was performed 186x744 times for each database.

Coding conditions			
JPEG images	Quality factors	T_0	
Images stored as features in	DB_1	$QF_{X_i} = 50$	IJG
	DB_2	$QF_{X_i} = 75$	IJG
	DB_3	$QF_{X_i} = 50$	HVS
Query images	$QF_Q = 40, 60, 85, 95$		IJG

Example of images (size 288x384)	IJG	HVS
	16 11 10 16 24 40 51 61 12 12 14 19 26 58 60 55 14 13 16 24 40 57 69 56 14 17 22 29 51 87 80 62 18 22 37 56 68 109 103 77 24 35 55 64 81 104 113 92 49 64 78 87 103 121 120 101 72 92 95 98 112 100 103 99	16 16 16 16 17 18 21 24 16 16 16 16 17 19 22 25 16 16 17 18 20 22 25 29 16 16 18 21 24 27 31 36 17 17 20 24 30 35 41 47 18 19 22 27 35 44 54 65 21 22 25 31 41 54 70 88 24 25 29 36 47 65 88 115

◆ Simulation result

◆ Measurement

$$TPR = \frac{TP}{TP+FN}, FPR = \frac{FP}{FP+TN}$$

TPR=100% means that there were **no false negative matches**.

Querying performance for images with $QF_Q = 60$

scheme	database	TPR[%]	FPR[%]
proposed	DB_1	100	0
	DB_2	100	0
FCS-based[1]	DB_1	100	0
	DB_2	0	0
image hashing[2]	DB_1	98.92	0.03
	DB_2	97.85	0.04

Querying performance for all query images

scheme	database	TPR[%]	FPR[%]
proposed	DB_1	100	0
	DB_2	100	0
	DB_3	100	0
FCS-based[1]	DB_1	75	0
	DB_2	50	0
	DB_3	71.23	0
image hashing[2]	DB_1	98.79	0.03
	DB_2	99.33	0.03
	DB_3	98.52	0.03

- ◆ **Only proposed scheme provided no false negative matches.**

- Image hashing-based one did in all cases.
- FCS-based one did under $QF_{X_i} < QF_Q$ or different T_0

Conclusion

- ◆ Our proposed identification scheme for JPEG images

- uses quantization matrices and the positions of zero values as features.
 - ⇒ They do not provide **no visible information**.
- uses the features and the property of DCT coefficients for identification.
 - ⇒ The use of them allows us to provide **no false negative matches in principle**.
- outperforms the querying performance.

Reference

- [1] K. Iida and H. Kiya, "Secure and Robust Identification Based on Fuzzy Commitment Scheme for JPEG Images," in *Proc. IEEE BMSB*, June, 2016.
- [2] Y. Li and P. Wang, "Robust image hashing based on low-rank and sparse decomposition," in *Proc. IEEE ICASSP*, March, 2017.