

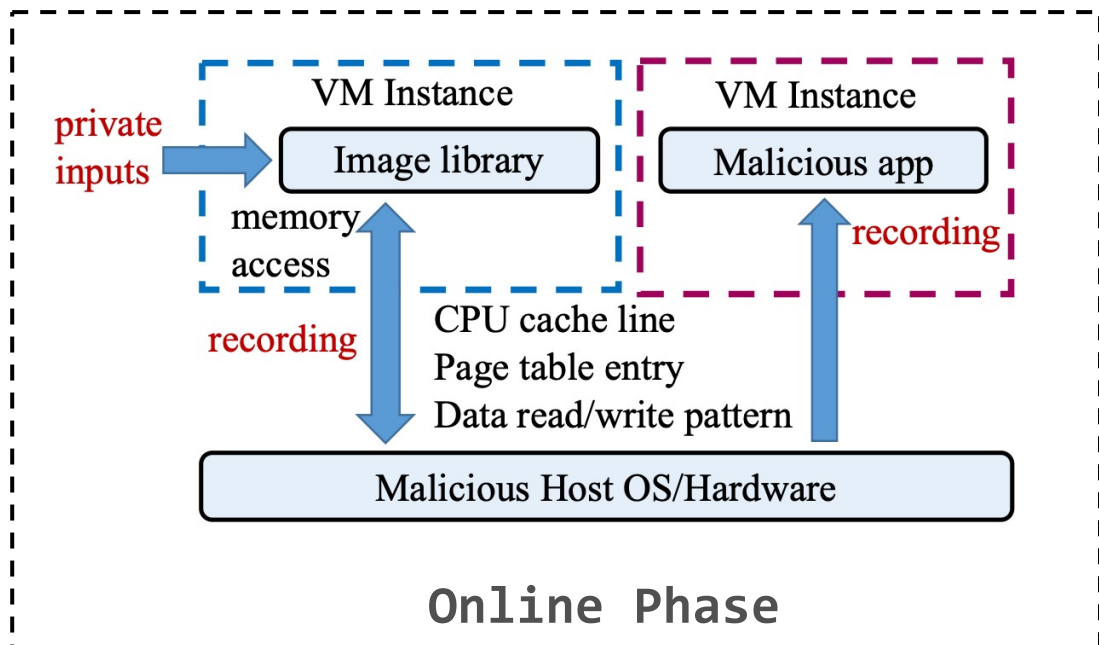
# Private Image Reconstruction from System Side Channels using Generative Models

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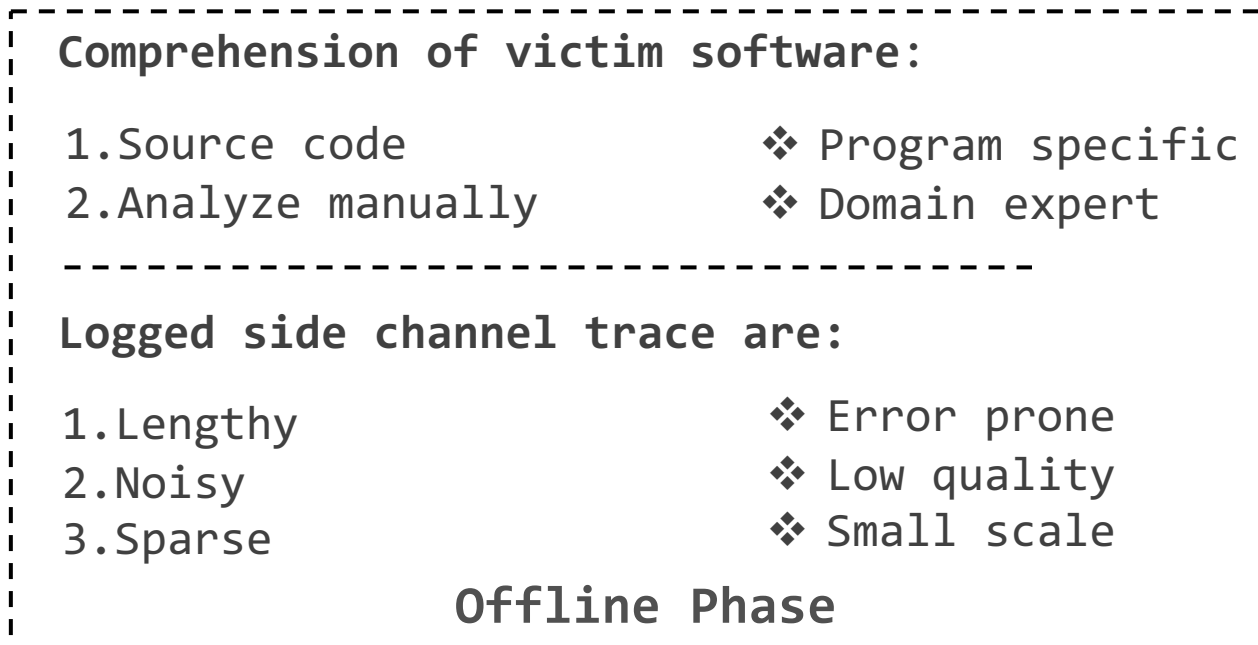
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# What is Side Channel Analysis?

*Recover privacy from program's non-functional characteristics*

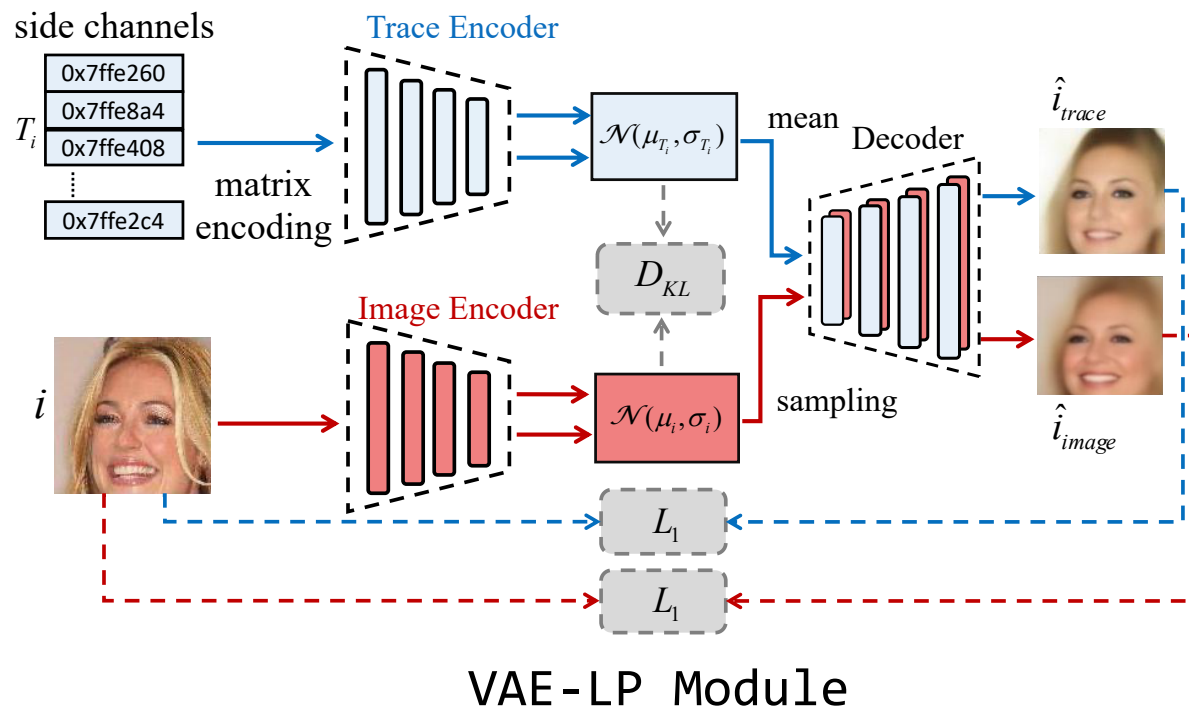


*Can be launched automatically!*



*Can be performed in an **effective** and fully **automated** way?*

# The Proposed Framework



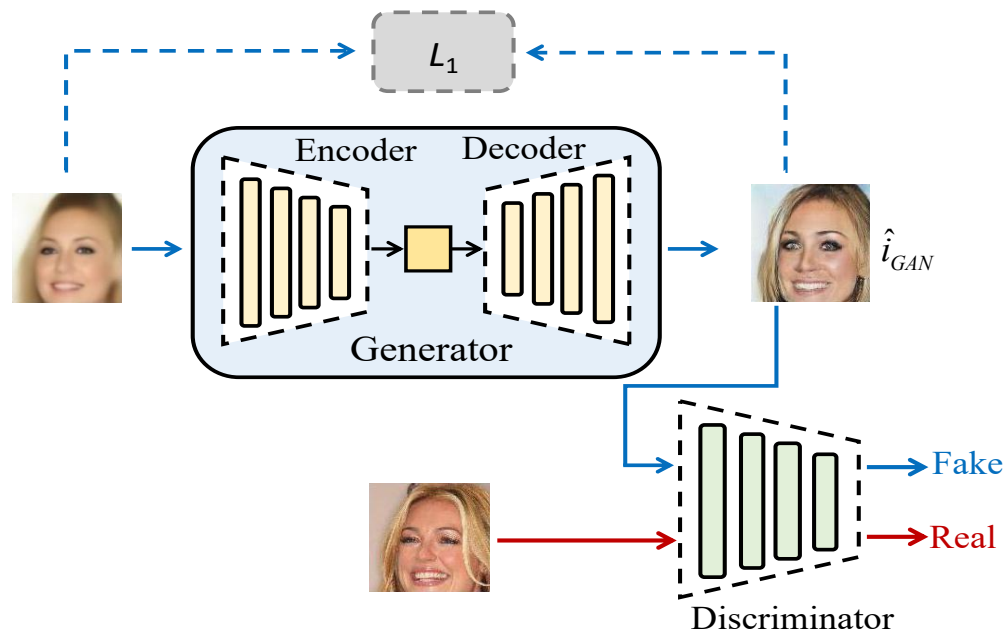
1. The upper dataflow extracts **secret-dependent records** from side channels.

2. The lower dataflow encodes and decodes **reference image**. The shared decoder works as a guideline.

Testing phase:

- Only retains the upper dataflow marked in **blue**.
- Side channels  $\rightarrow$  Private images

# The Proposed Framework

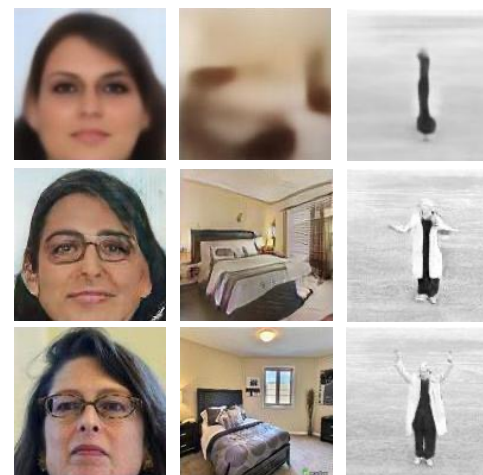


## GAN Module

Accepts output from VAE-LP module and refine it.

Why refine?

Human adversaries!

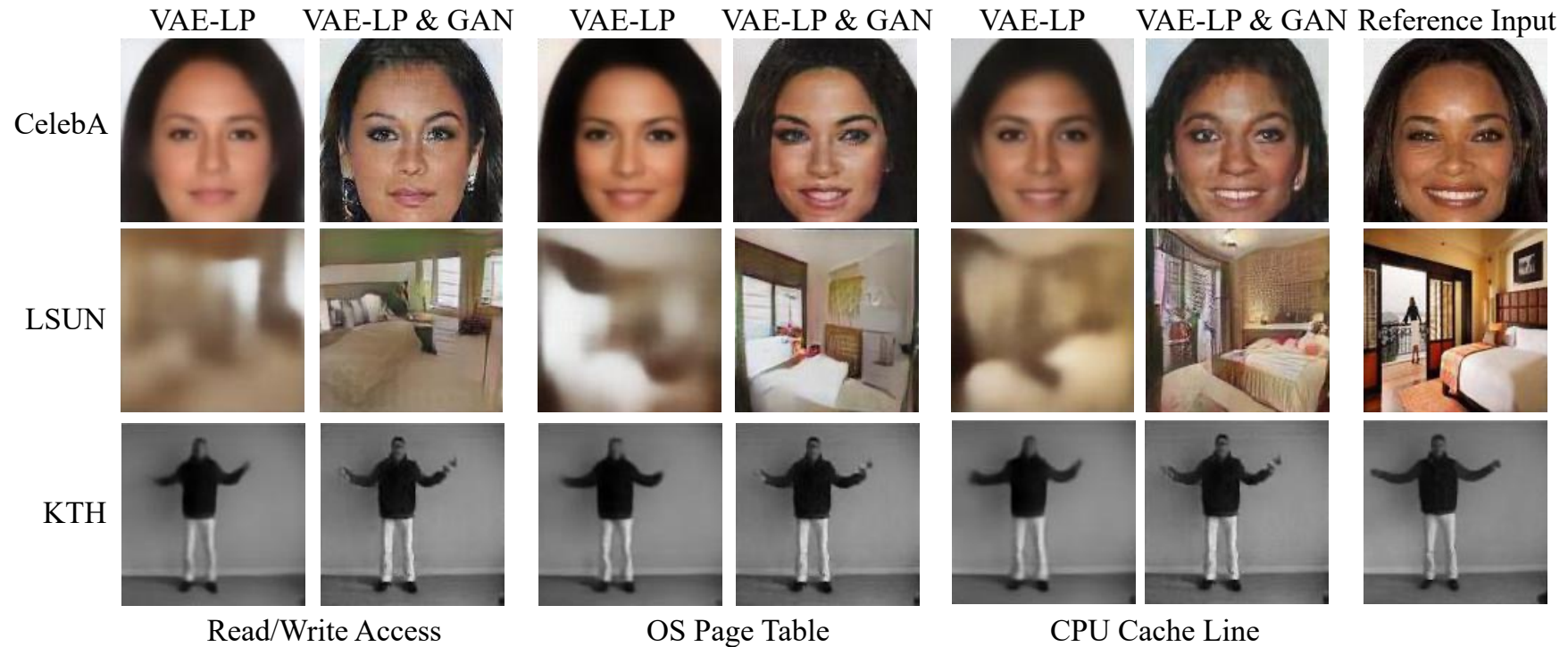


VAE-LP Module

GAN Module

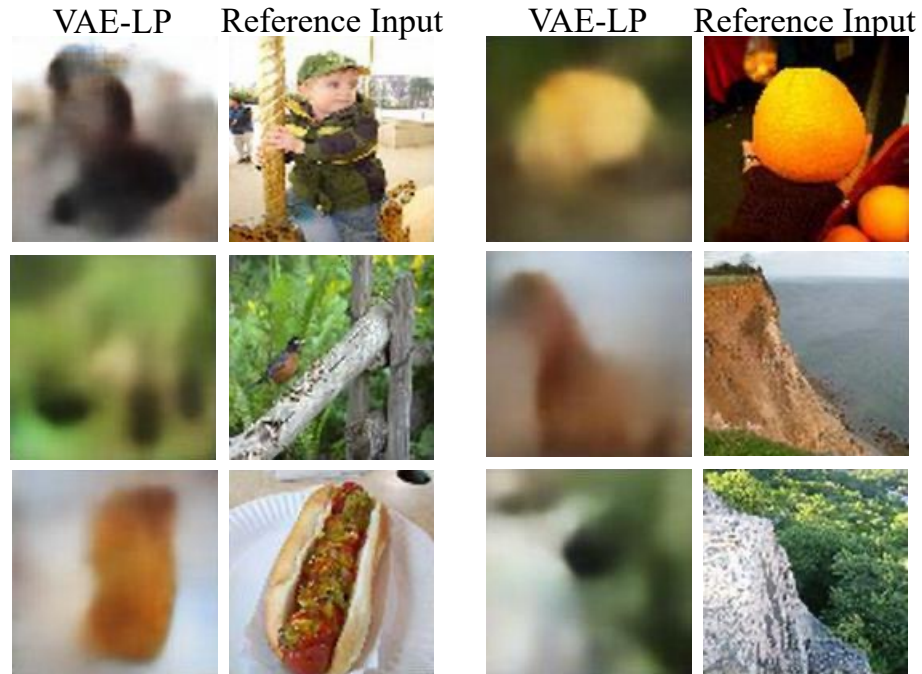
Reference Input

# Qualitative Evaluation



Reconstructed images on three datasets using three side channels.

# Qualitative Evaluation



1. Images are mixed
2. No class label is provided

Reconstructed images on mini-ImageNet using CPU cacheline.

# Quantitative Evaluation

System side channel	$K = 1, N = 100$			$K = 5, N = 100$			$K = 20, N = 100$		
	CelebA	KTH	LSUN	CelebA	KTH	LSUN	CelebA	KTH	LSUN
CPU Cache Line	47.28%	38.22%	10.00%	74.98%	63.56%	28.74%	90.70%	83.72%	58.88%
OS Page Table	16.16%	35.86%	2.26%	37.96%	61.28%	8.62%	65.14%	82.22%	26.42%
Read/Write	19.58%	33.44%	2.24%	42.44%	58.30%	8.28%	67.82%	80.14%	26.64%

mini-ImageNet/(K, N)	(1, 100)	(5, 100)	(20, 100)
CPU Cache Line	18.8%	40.6%	66.0%

Given  $N$  images, whether reference input is in the top- $k$  similar images using SSIM score.