

From Noise to Readable Images: Eavesdropping on Computer Screens via Custom Hardware and Deep Learning Image Reconstruction

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Image Transmission

A computer's image is sent to the monitor as a stream of RGB pixels.

- pixel's R, G and B values are mapped to voltages,
- these voltages are transmitted through a cable,
- in total, each second $x_t \cdot y_t \cdot f_v$ pixels are transmitted,
 - x_t – display width, y_t – display height, f_v – display frame rate.

- rapid changes in voltages → electromagnetic emissions,
 - can be captured from distance with a software defined radio (SDR),
- pixel transmission period $t_p = 1/(x_t \cdot y_t \cdot f_v)$,
 - impulses in the captured signal at multiples of t_p ,
 - impulse amplitude → grayscale value of the pixel,
- automatic reconstruction with TempestSDR,
 - open-source software, developed by Martin Marinov,
 - offers real-time image reconstruction.

TempestSDR Example Attack

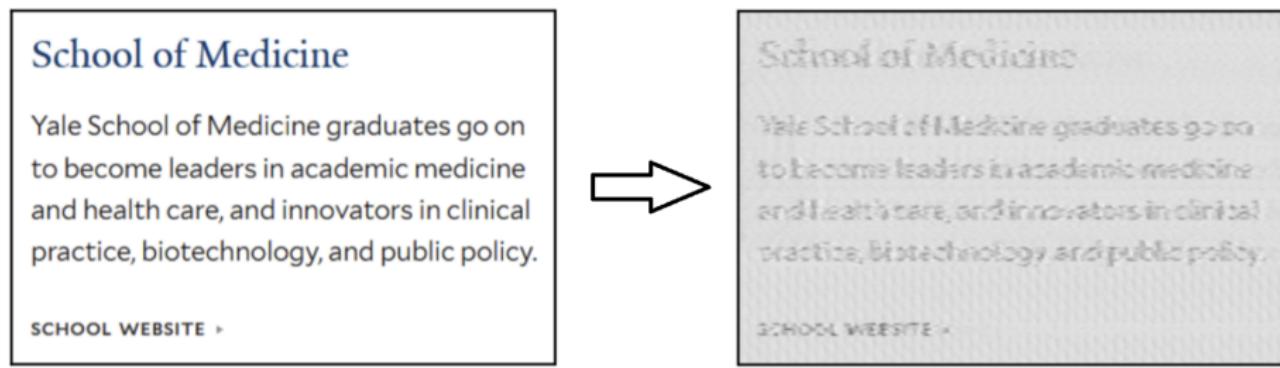


Figure 1: Left to right: original image, reconstructed image in TempestSDR.

Custom Hardware

With standard hardware, the attack range is limited to 1 meter.

- ① 5-element Yagi-Uda antenna tuned to 130 MHz,
- ② low-noise amplifier for signal gain,
- ③ band-pass filter (125-135 MHz) for signal isolation.

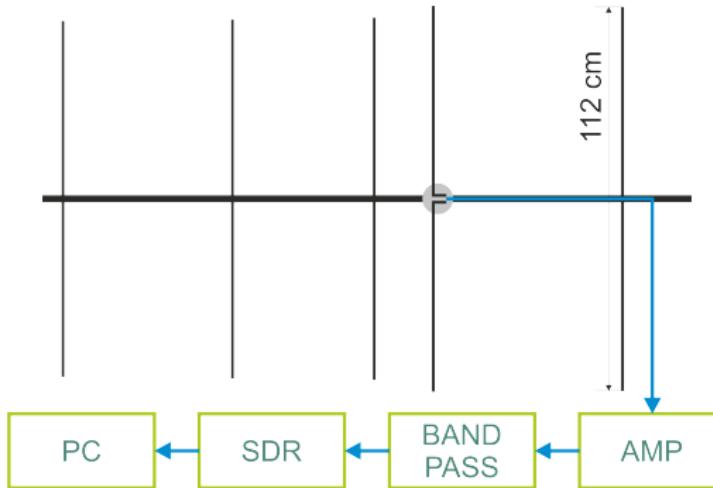


Figure 2: Our Yagi-Uda system.



Figure 3: Reference image.

Custom Hardware

Tested in a hallway with a metal structure.

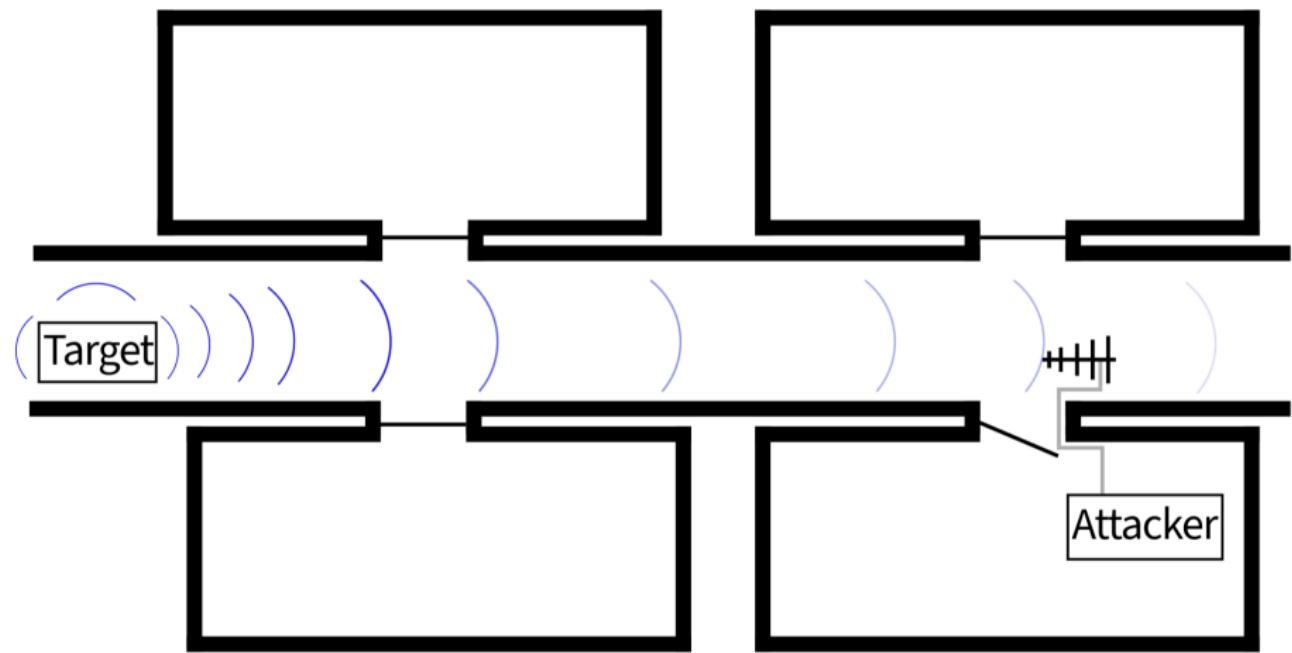


Figure 4: Attack scenario.

Custom Hardware



Figure 5: Captured image from various distances.

Deep Learning Image Reconstruction

Captured images still remain noisy.

- we created a dataset of common computer usages,
 - web pages, text documents, OS interfaces, ...
 - 600 real images + 250 generated images,
 - each image captured ~ 6 times with TempestSDR,
 - automated with PyAutoGUI, Selenium, OpenCV,
- trained two deep learning models for image reconstruction,
 - auto-encoder (Deep Residual U-Net, DRUNet),
 - noise-predictor (Denoising CNN, DnCNN).

Deep Learning Image Reconstruction



Figure 6: An illustrative sample from our dataset.

Deep Learning Image Reconstruction

- PSNR (Peak Signal-to-Noise Ratio),
- SSIM (Structural Similarity Index Measure),
- CER (Character Error Rate).

Method	PSNR	SSIM	CER
TempestSDR	8.65 dB	0.41	86%
DnCNN	17.63 dB	0.75	54%
DRUNet	19.08 dB	0.87	46%

Table 1: Image quality comparison of captured (TempestSDR) and reconstructed (DRUNet and DnCNN) images.

Image Reconstruction Example



Figure 7: Example of the attack.

Image Reconstruction Example



Figure 8: Example of the attack.

Image Reconstruction Example

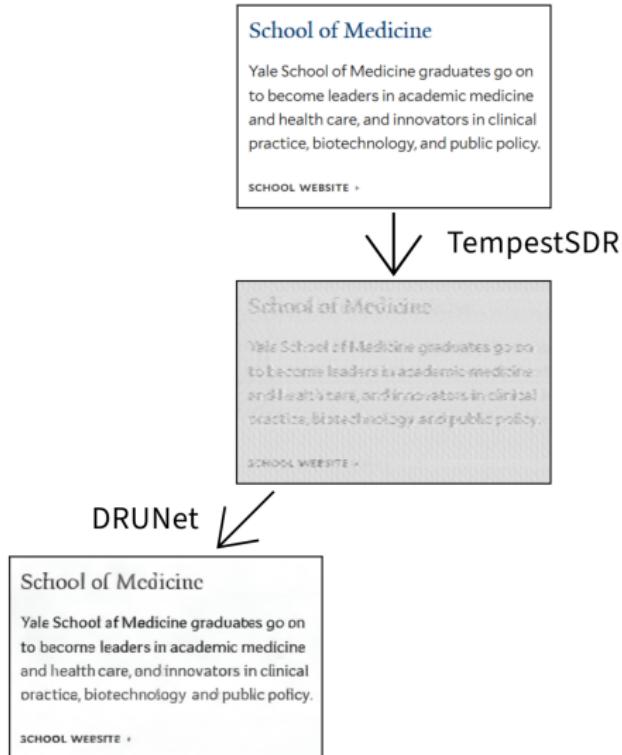


Figure 9: Example of the attack.

Image Reconstruction Example

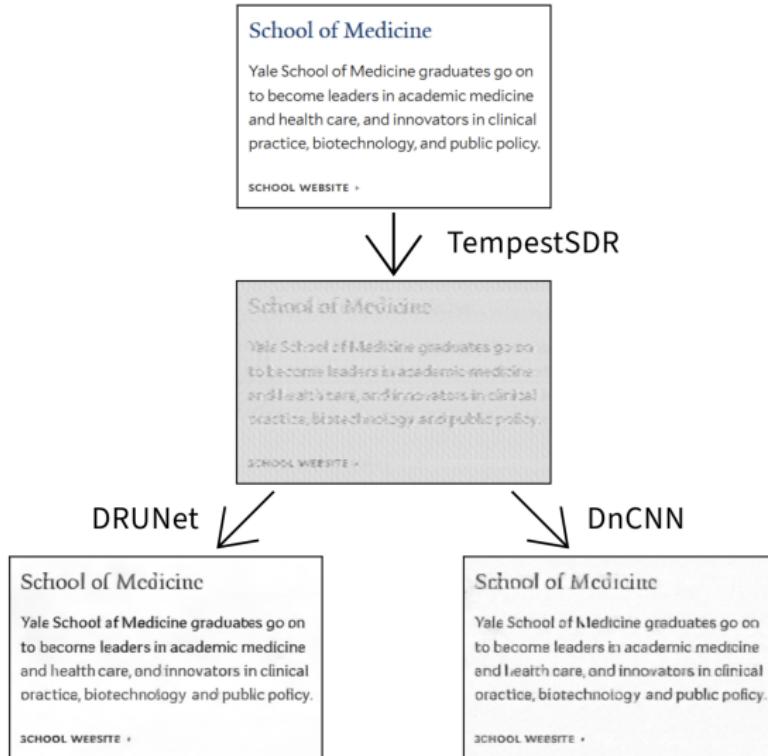


Figure 10: Example of the attack.

Thanks for your attention!