

Image Watermarking

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What is Watermark?

Protect Copyrights!

What is Watermark

- Digital Watermark is the technology to leave the copyright information in an image itself



Steganography VS Watermark

- Watermark
 - protect copyright of the image
 - publisher leave a message to an image to protect the rights
- Steganography
 - send secret message to receiver
 - sender and receiver want to communicate without leaking to adversary

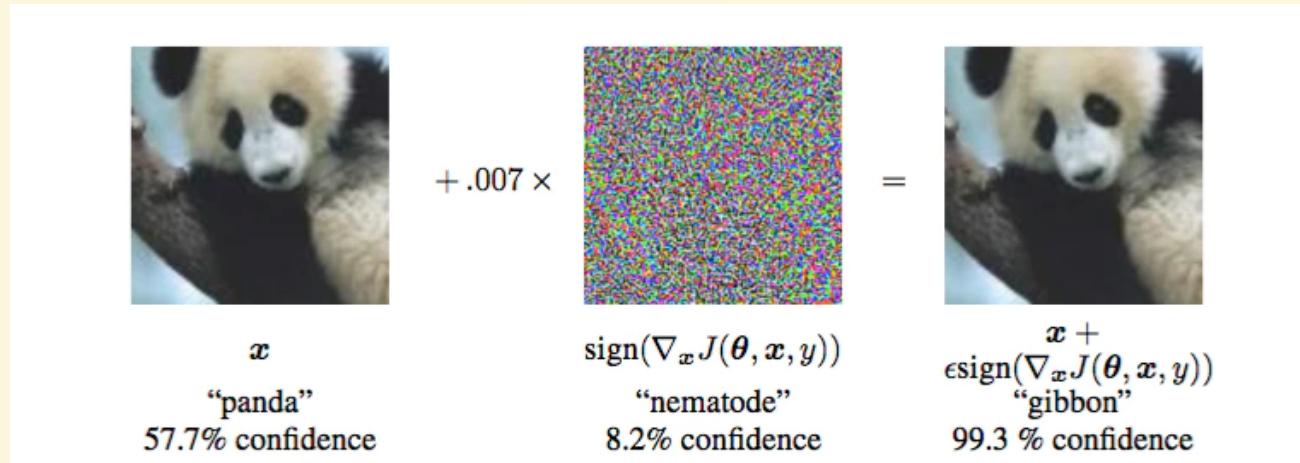
Visible and Invisible Watermark

- **Visible**
 - Ordinary Watermark which often used
 - Easy to put it
- **Invisible**
 - Human cannot recognize watermark
 - Put watermark in the imperceptible space to human



Why Watermark with NN?

- NN is sensitive to tiny perturbation in image ([GoodFellow, 2014](#))
 - This feature cause adversarial attack problem which cannot be solved yet
- We can use this sensitivity to hide and extract information in image



Invisible Watermark

You cannot see it!

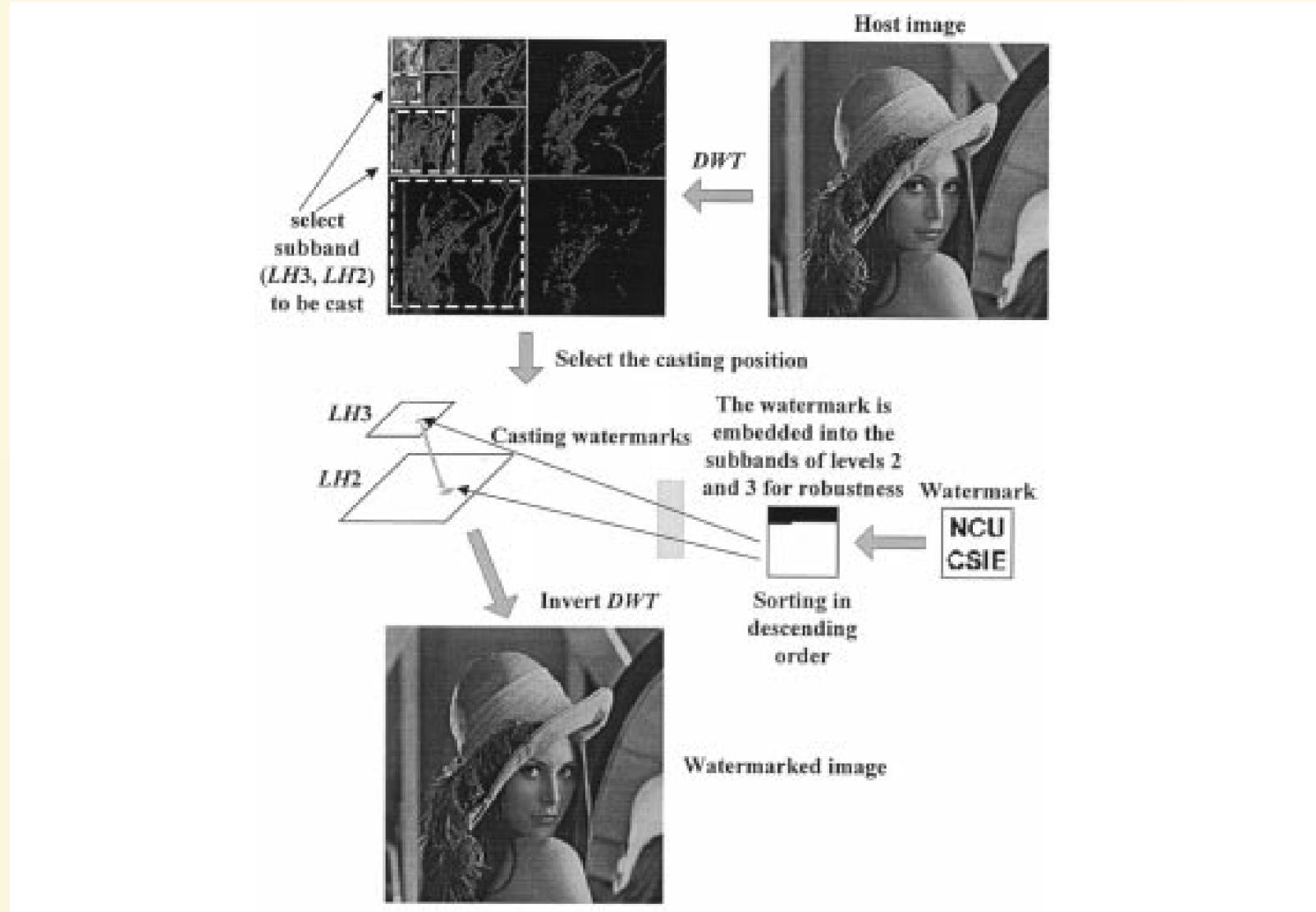
Traditional Methods

Fourier Transformation

- Robust to linear transformation ([Shelby, 2000](#))

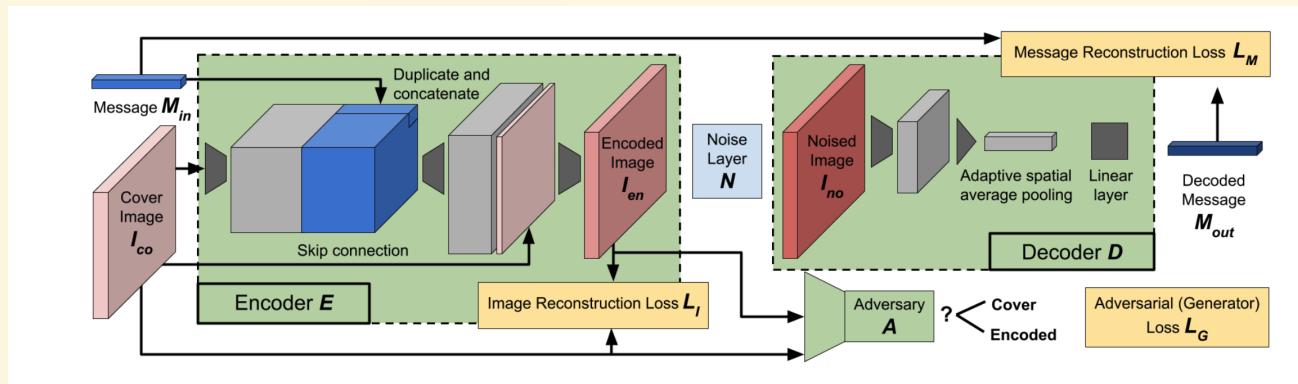
Wevelet Transformation

- Two-band wavelet transform, embed the watermark bits on the wavelet coefficients ([Ning, 2007](#))
- Multiband one, embedd it in the mean trend of some middle-frequency subimages ([Ming, 2001](#))

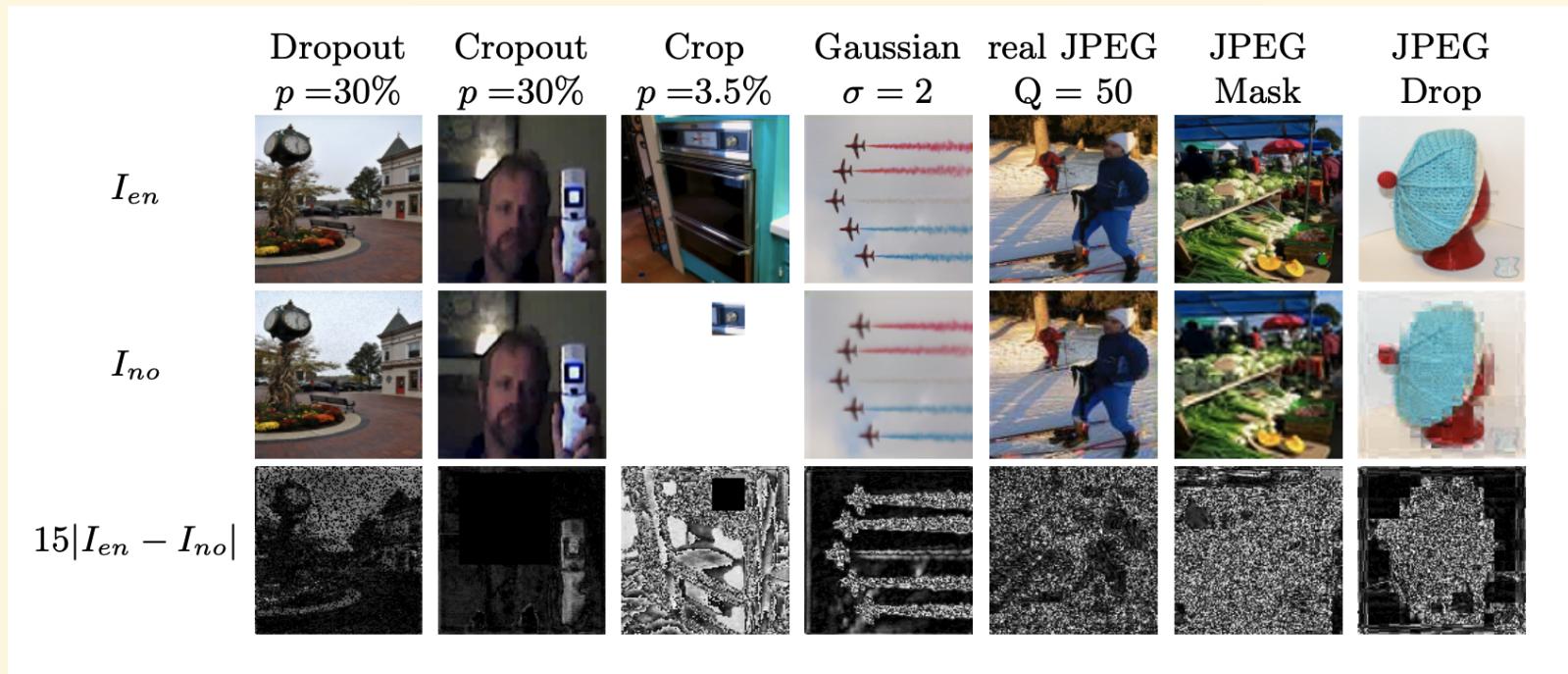


HiDDeN: Hiding Data with Deep Networks

- More robust better image quality than传统s
- Model
 - **Encoder**: encode message to image
 - **Adversary**: Distort the encoded image
 - gaussian blur, dropout, crop, jpeg compression, ...
 - **Decoder**: Decode the encoded message from the image

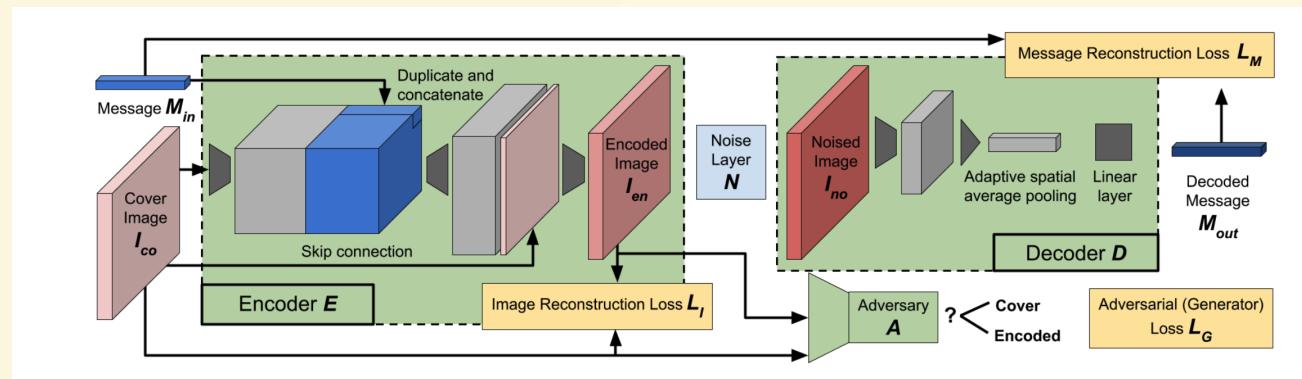


Adversary (Noise)



Loss Function

- 3 Loss functions
 - Image Reconstruction Loss
 - Difference between cover and encoded image
 - Message Reconstruction Loss
 - Difference between original and decoded message
 - Adversarial Loss
 - Adversarial Loss for cover and encoded image



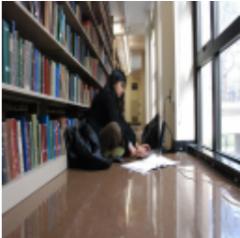
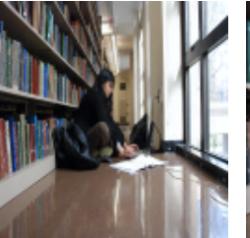
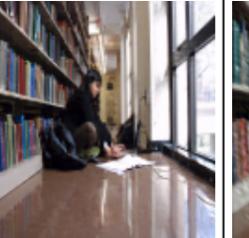
Metrics

- Capacity
 - Bits per Pixel ($= L / HWC$)
- Secrecy
 - PSNR (larger is higher quality)
 - Detection Rate (for steganography)
- Robustness
 - Bit Accuracy of Message

Important metrics for steganography and watermark are different
Robustness is most important for watermark

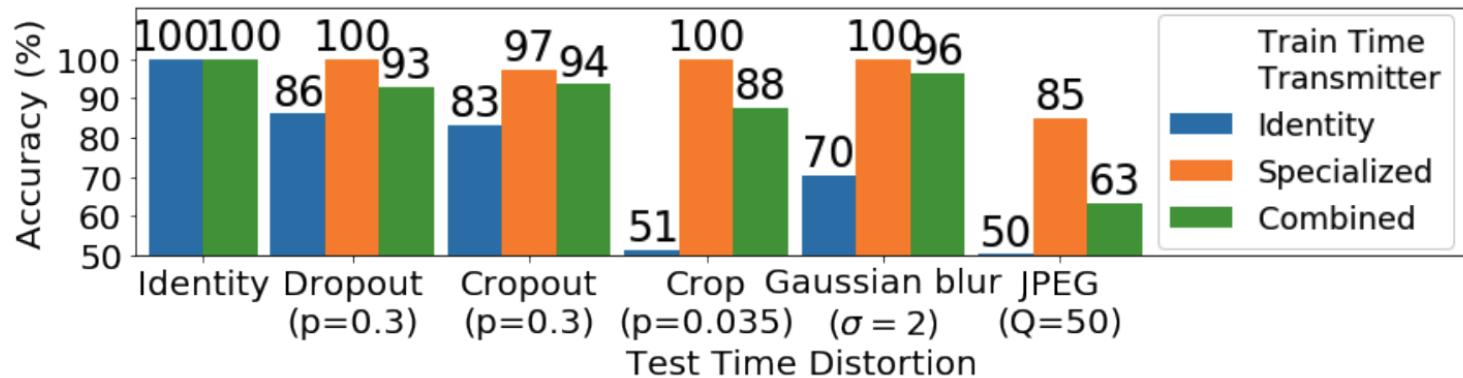
Experiment / Secrecy

- Quality is not bad

	Digimarc	Identity Dropout	Cropout	Crop	Gaussian	JPEG-mask	JPEG-drop	Combined
PSNR(Y)	62.12	44.63	42.52	47.24	35.20	40.55	30.09	28.79
PSNR(U)	38.23	45.44	38.52	40.97	33.31	41.96	35.33	32.51
PSNR(V)	52.06	46.90	41.05	41.88	35.86	42.88	36.27	33.42
Cover		Trained with Adversary						No Adversary
Cover	Digimarc	Crop	Gaussian	Combined	Combined			
								

Experiment / Robustness

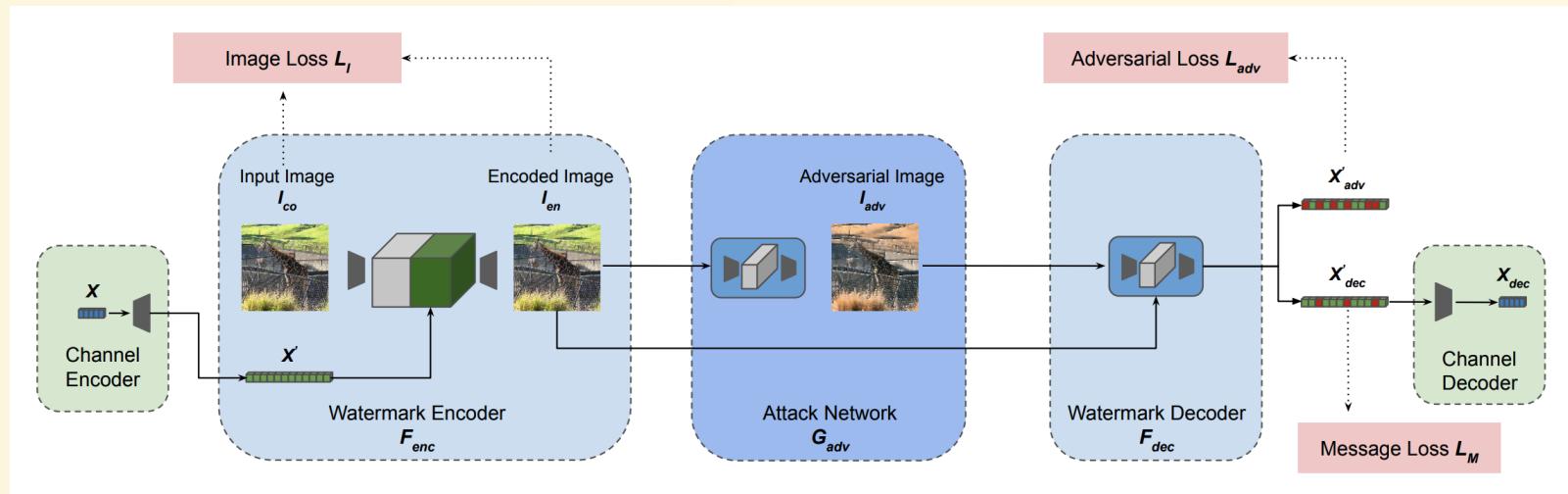
- JPEG is hardest distortion
- Specialized trained model is robust except for JPEG
- Non-adversary trained model cannot deal with Crop



Distortion Agnostic Deep Watermarking

CVPR2020

- "Attack Network" distort the encoded image with NN
 - Not define the specific distortion type
- "Channel Encoding" treat the message as the feature vector in the model



Result



Original



Encoded



Difference

Figure 2: Example of original image, encoded image and difference between the two images from our model.

- As robust as the specialized and combined of HiDDeN

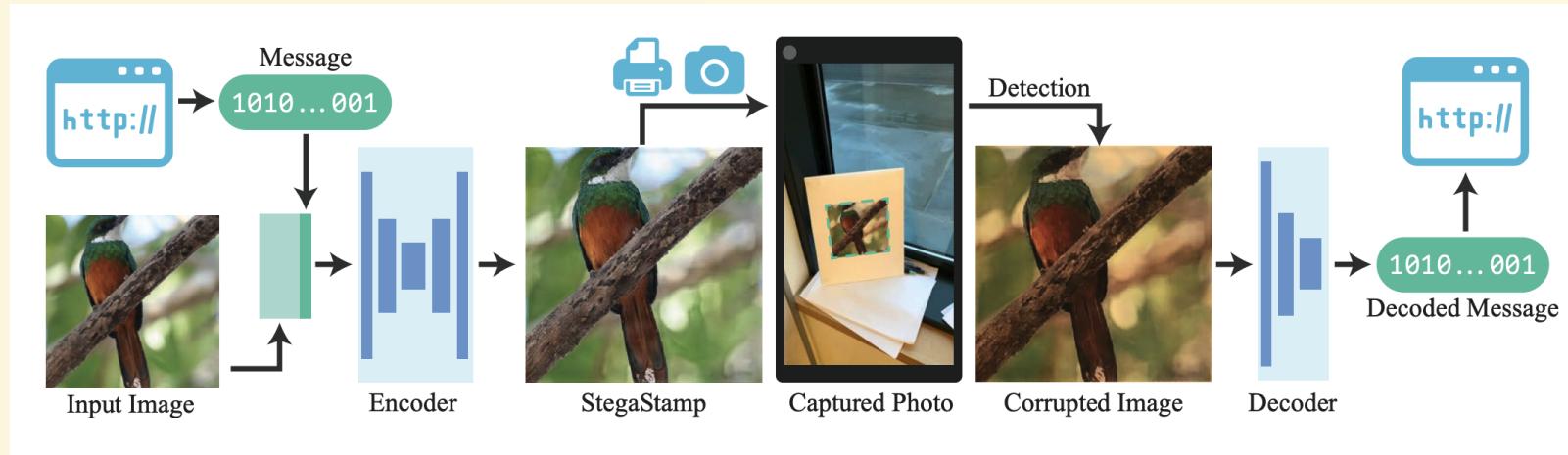
Watermarking Models	Known Distortions					Unknown Distortions				
	Identity	50.2	53.0	81.3	71.8	74.6	65.1	93.4	85.3	65.0
JPEG (Q=50)	99.0	80.8	85.7	87.0	86.1	87.7	71.0	95.2	78.7	90.6
Crop ($p=0.035$)	100.0	49.9	99.3	51.0	62.5	50.3	64.3	92.1	81.1	63.3
Dropout ($p=0.3$)	93.1	50.0	51.3	99.4	51.0	51.0	54.3	89.0	75.4	63.8
Blur ($\sigma=1.0$)	53.0	49.8	51.8	50.5	99.9	50.3	61.4	69.6	76.3	52.6
<i>Combined [40]</i>	100.0	77.0	99.1	98.7	99.1	93.5	70.8	94.2	84.9	88.6
Our Model	100.0	81.7	93.5	97.9	92.8	95.6	94.0	98.5	88.4	91.7

Identity
 JPEG (Q=50)
 Crop ($p=0.035$)
 Dropout ($p=0.3$)
 Gaussian Blur ($\sigma=1.0$)
 Gaussian Noise ($\sigma=0.06$)
 Hue ($\delta=0.2$)
 Saturation (15.0)
 Resize (0.7)
 GIF (P=16)

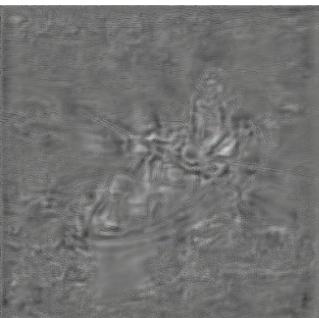
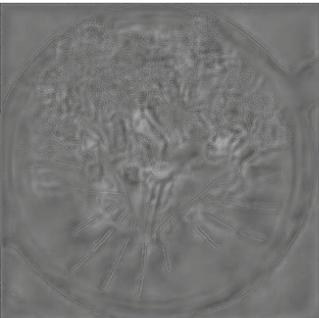
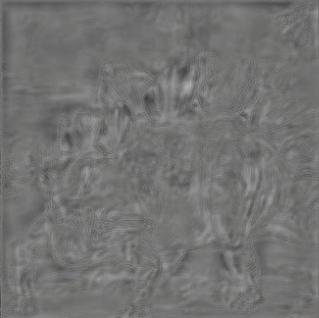
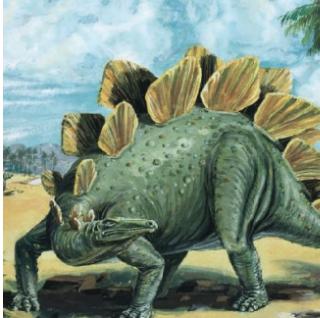
StegaStamp: Invisible Hyperlinks in Physical Photographs

CVPR2020

- Realize the traceability of realworld phisycal images
- e.g. Embed URL to realworld images, and it can be scanned by AR glass



Encoded images

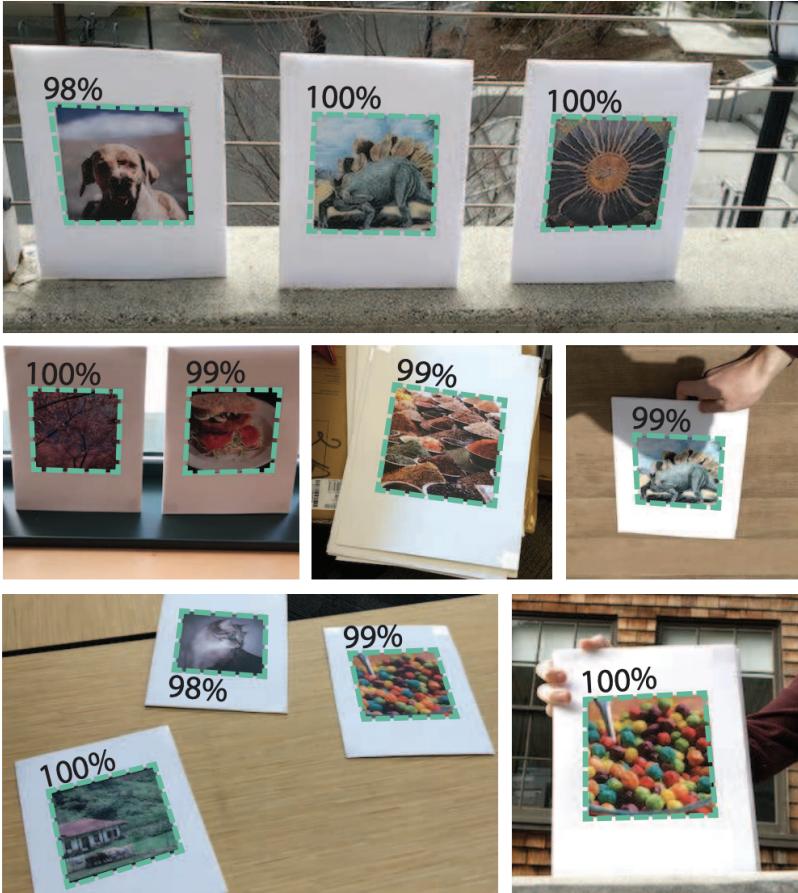


Original Image

StegoStamp

Residual

Realworld Experiment

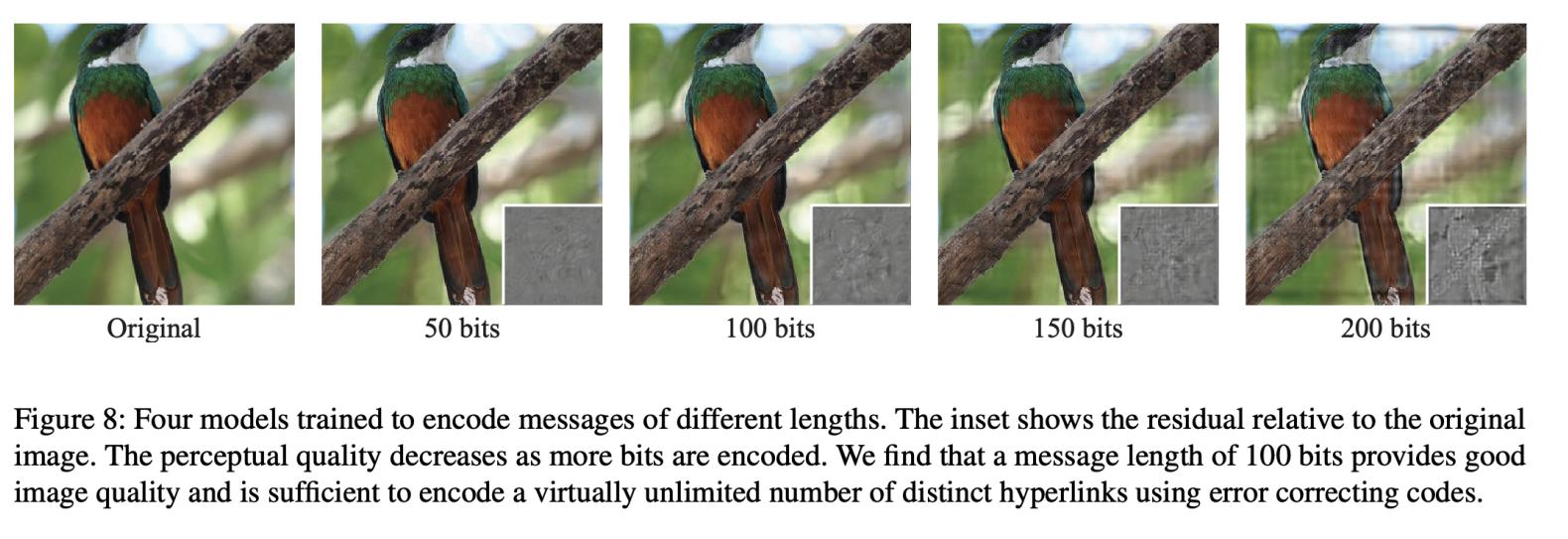


Decoding Accuracy

		5th	25th	50th	Mean	
Webcam	Printer	Enterprise	88%	94%	98%	95.9%
	Screen	Consumer	90%	98%	99%	98.1%
	Printer	Pro	97%	99%	100%	99.2%
	Screen	Monitor	94%	98%	99%	98.5%
	Printer	Laptop	97%	99%	100%	99.1%
	Screen	Cellphone	91%	98%	99%	97.7%
Cellphone	Printer	Enterprise	88%	96%	98%	96.8%
	Screen	Consumer	95%	99%	100%	99.0%
	Printer	Pro	97%	99%	100%	99.3%
	Screen	Monitor	98%	99%	100%	99.4%
	Printer	Laptop	98%	99%	100%	99.7%
	Screen	Cellphone	96%	99%	100%	99.2%
DSLR	Printer	Enterprise	86%	96%	99%	97.0%
	Screen	Consumer	97%	99%	100%	99.3%
	Printer	Pro	98%	99%	100%	99.5%
	Screen	Monitor	99%	100%	100%	99.8%
	Printer	Laptop	99%	100%	100%	99.8%
	Screen	Cellphone	99%	100%	100%	99.8%

Encoded images for different message length

- To embed longer message, some square artifacts appear

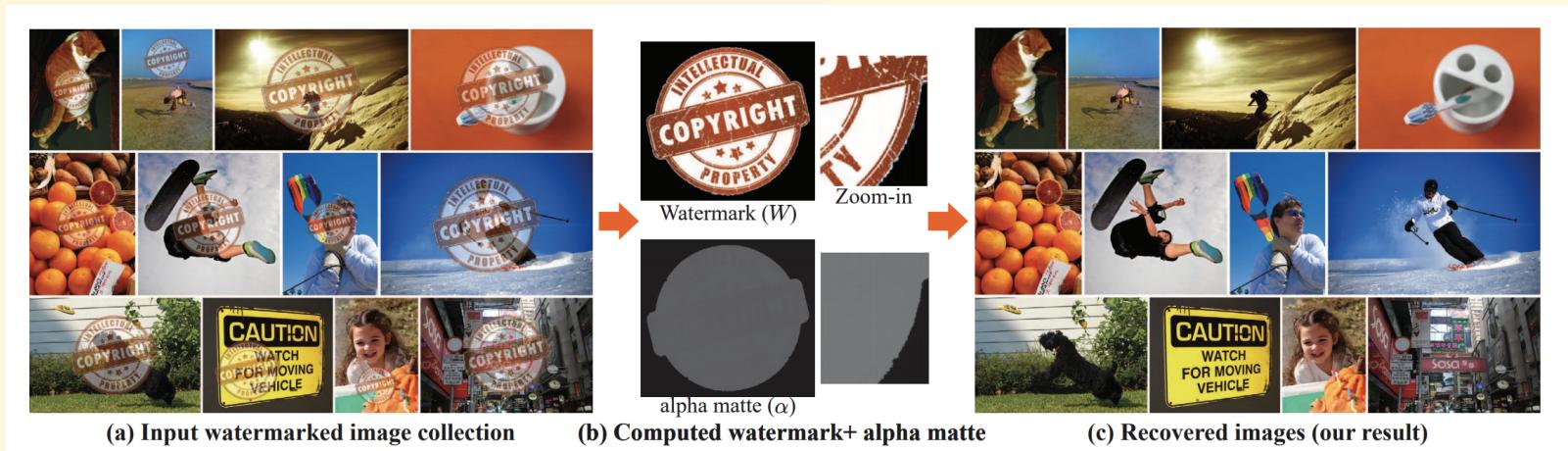


Visible Watermark Removal

Is it really effective?

On the Effectiveness of Visible Watermarks

- Easy to remove visible watermark by exploiting collection of watermarked images
- This is image collection approach, it's new



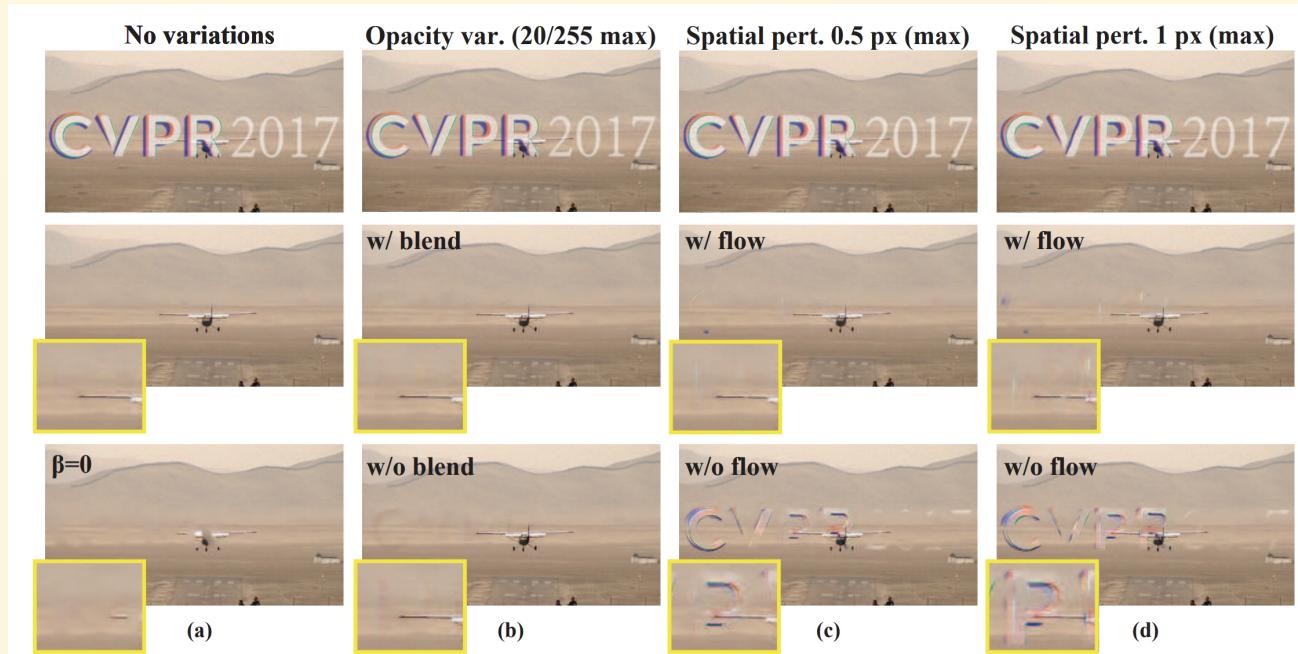
Comparison to baseline

- Our result is best quality



Analysis for inconsistency watermark pattern

- Not robust to spatial perturbation
- Do perturbation affect to reconstruction phase?



Result for stock imagery

Datasets	# Images	Long Edge	Watermark Res.(w×h)	Pipeline (min)	Removal (sec)
<i>AdobeStock</i>	422	1000	623 × 134	33	18
<i>123RF</i>	1376	650	650 × 433	115	60
<i>CanStock</i>	3000	450	450 × 300	28	12
<i>fotolia</i>	285	500	199 × 66	5	1.5
<i>CVPR17</i>	1000	640	403 × 67	34	5
<i>Copyrights</i>	1000	640	339 × 307	47	40

AdobeStock (422 images), c=0.41



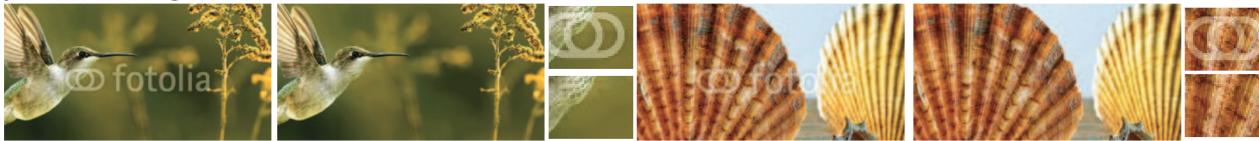
123RF (1340 images), c=0.2



CanStock (3000 images), c=0.17



fotolia (285 images), c=0.45



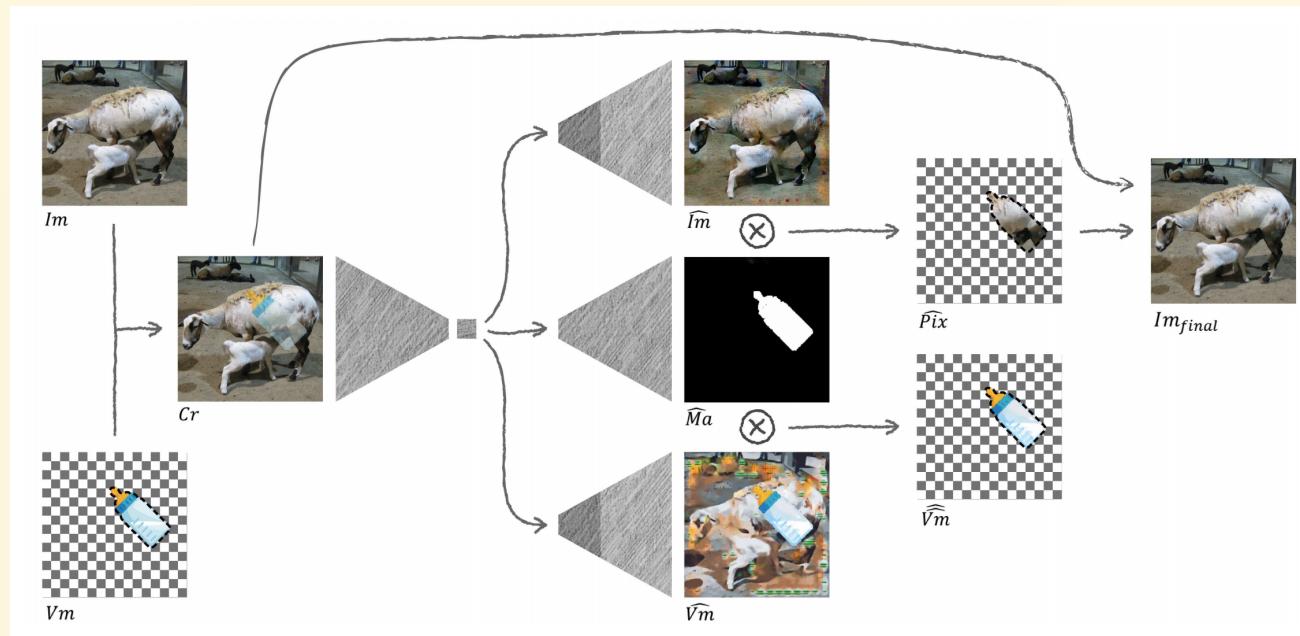
Blind Visual Motif Removal from a Single Image

- Remove visual motif (such as watermark) from image by using only single image
- Easy to remove
- This is blind removal



Method

- Extract background, motif area, foreground (motif) from the image, and then reconstruct the original image



Result images



Appendix: Model Watermarking

Protect copyrights of your model!

Coming soon!

Summary

Invisible Watermark with NN

- Under development
- Encoder / Adversary / Decoder framework is the best
- Robustness is most important

Visible Watermark Removal

- Single Image VS Multiple Image (Blind VS Non-Blind)
- Strong watermark which cannot be removed is needed