

# Stego network

# Steganography

- *“The best way of keeping a secret is to pretend there isn’t one.”*— Margaret Atwood
- *“Steganography is the art and science of hiding communication.”*

# Steganography

- Usually
  - Stego medium is image, “stego image”
  - Secret data can be image, text, and etc.
- Three elements
  - Capacity: 얼마나 많이 cover medium 에 숨길 수 있는가?
  - Security: 도청자에게 들키지 않을 수 있는가?
  - Robustness: Cover medium 이 수정될 때 숨겨진 내용을 보존할 수 있나?

# Steganography

- Steganography
  - Invisible
  - It is okay to destroy if the stego medium is modified
  - Our interest and goal
- Watermarking
  - Visible / invisible
  - It should be retained while the stego medium is modified
  - Copyright ...

# Steganography: example

- MSB on LSB
  - The simplest method
  - Visual attack
- DCT coefficient based method
  - Use LSB of DCT coefficient
  - No visual attack
  - Steganalysis can be applied

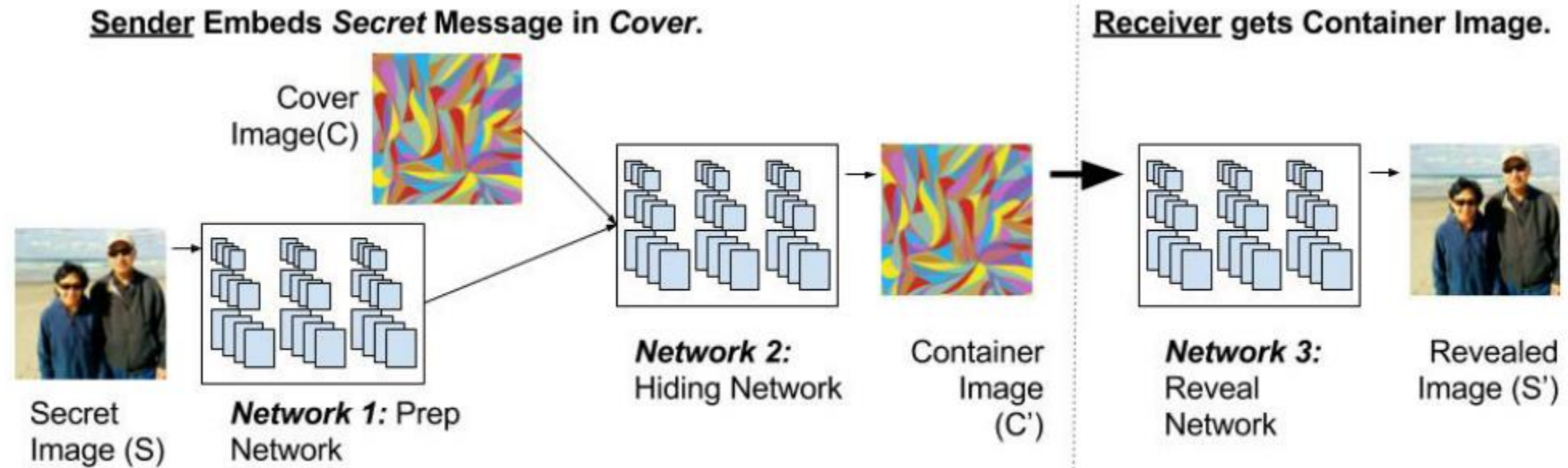
# Idea

- If stego medium is a neural network
  - We cannot understand parameter itself by looking at
  - Pixel is an integer value, but the parameter is real value
  - It is hard to apply steganalysis technique because we do not know the nature of NN parameters
  - Neural network can inference data
    - We need to consider parameter, activation, output, and gradients
    - Accuracy for discriminative task (regression, classification)
    - Quality of outputs for generation task (GAN, VAE)

# Related work

- No exact one
- Hiding images in plain sight: Deep steganography (NIPS 2017)
- HiDDeN: Hiding data with deep networks (ECCV 2018, Justin Johnson / Li Fei-Fei)

# Deep steganography





# Deep steganography

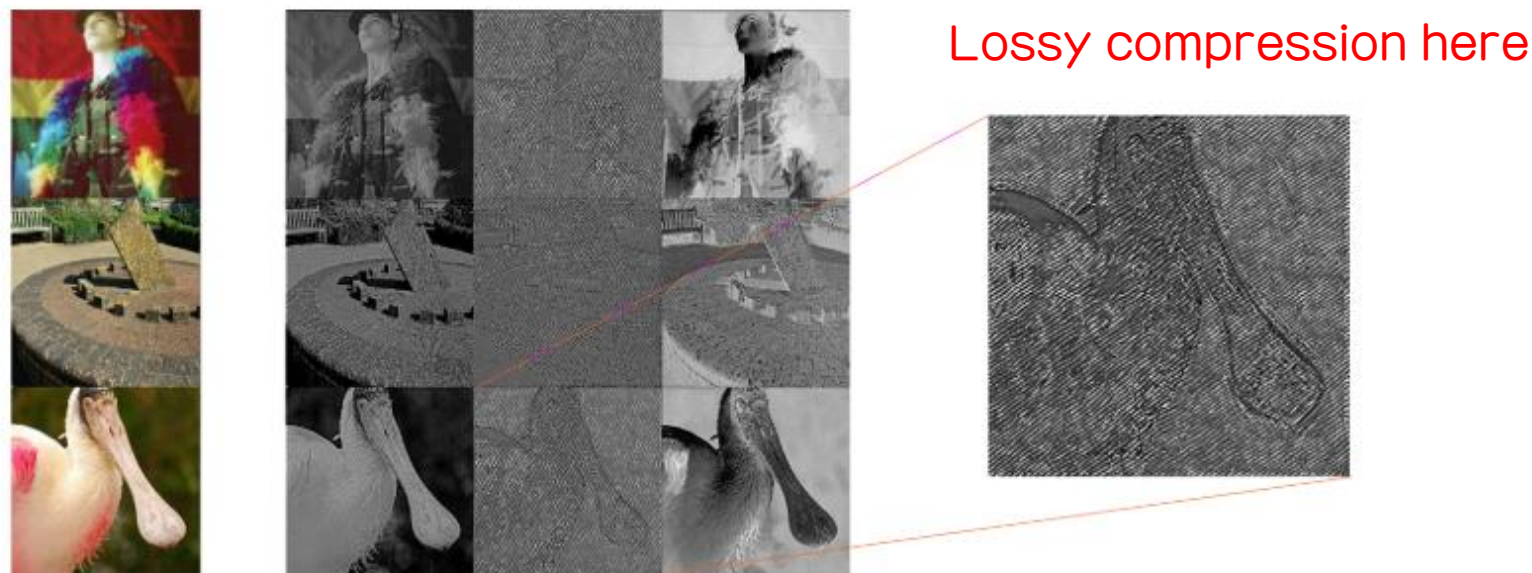


Figure 2: Transformations made by the preparation network (3 examples shown). Left: Original Color Images. Middle: the three channels of information extracted by the preparation network that are input into the middle network. Right: zoom of the edge-detectors. The three color channels are transformed by the preparation-network. In the most easily recognizable example, the 2nd channel activates for high frequency regions, e.g. textures and edges (shown enlarged (right)).

# Deep steganography

$$L(c, c', s, s') = ||c - c'|| + \beta ||s - s'||$$

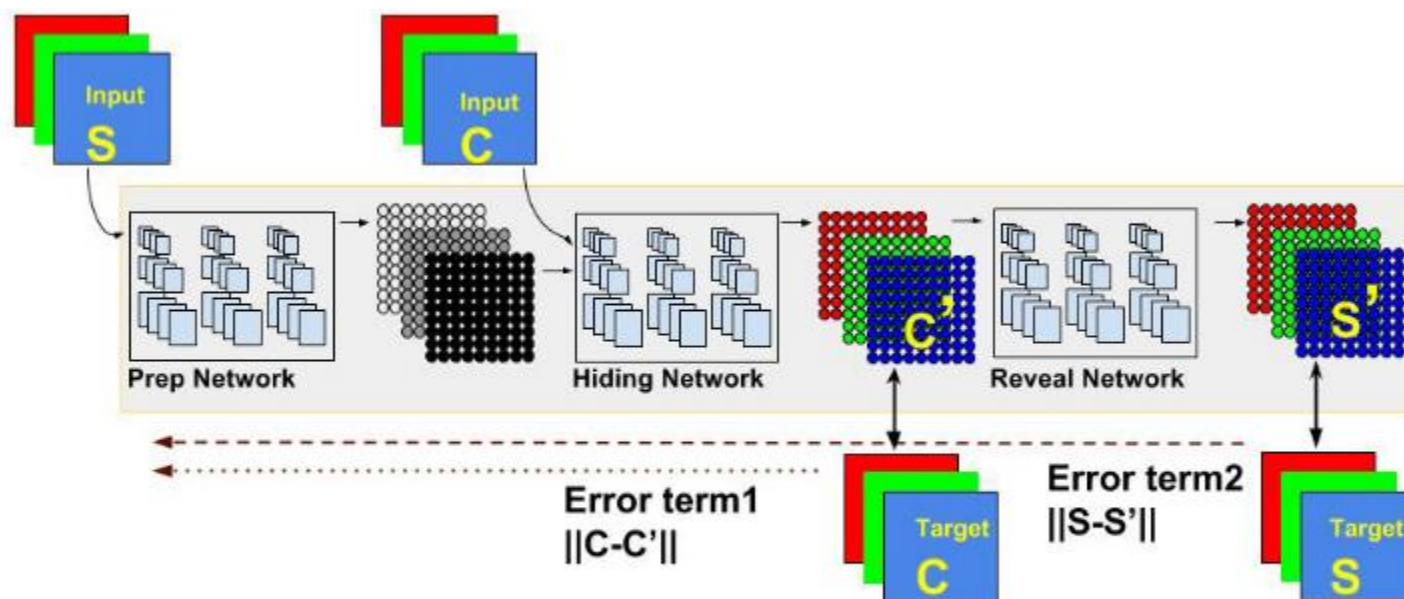
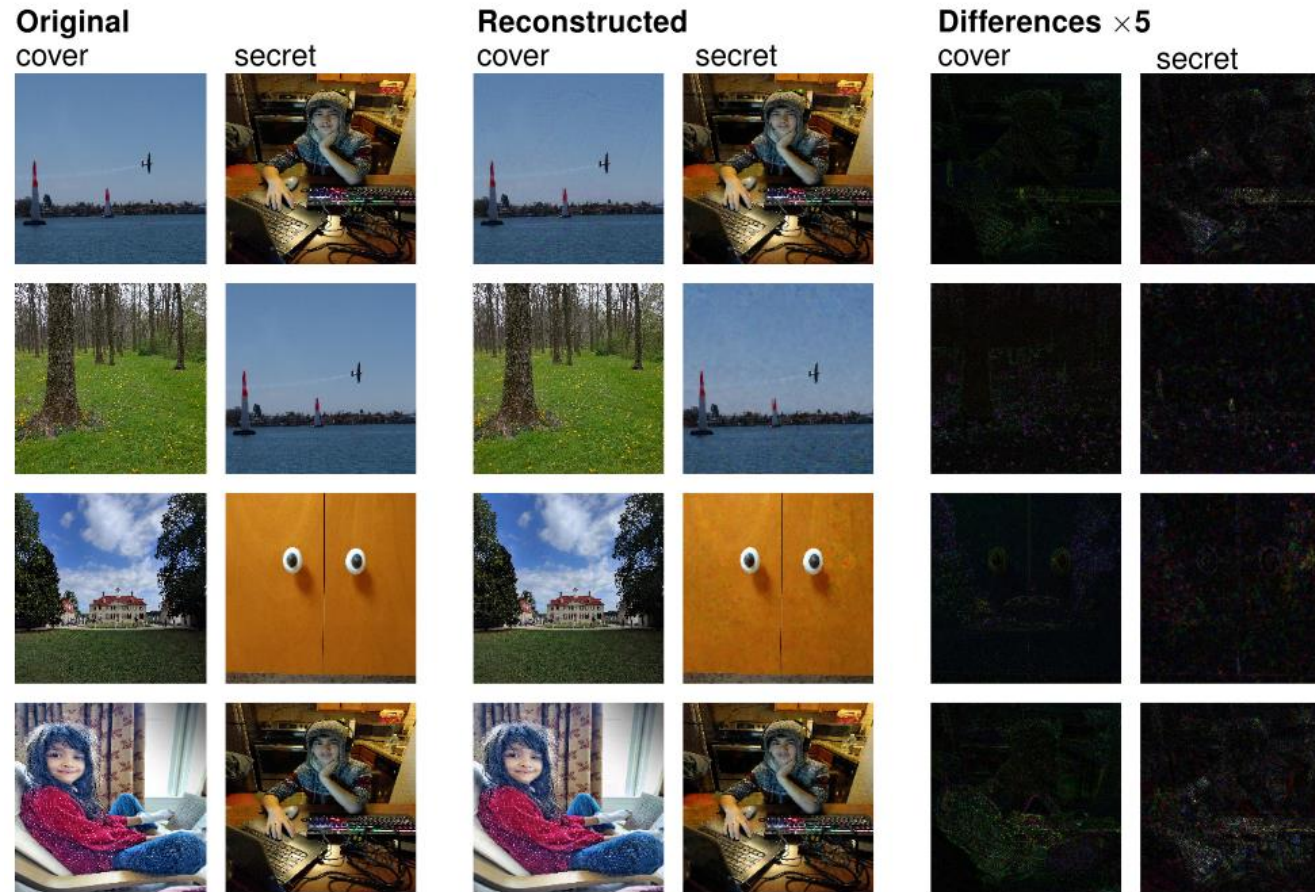


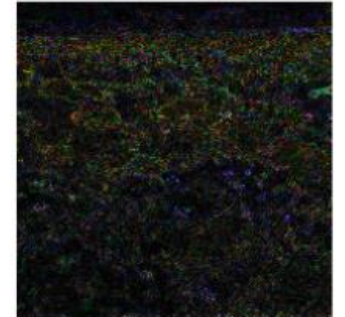
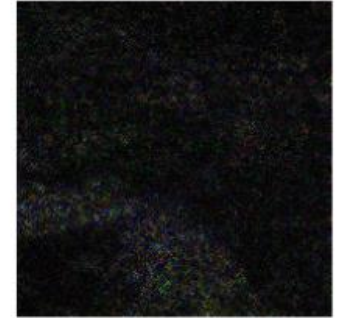
Figure 3: The three networks are trained as a single, large, network. Error term 1 affects only the first two networks. Error term 2 affects all 3.  $S$  is the secret image,  $C$  is the cover image.

# Deep steganography





# Deep steganography



# Deep steganography

**Original  
Cover**



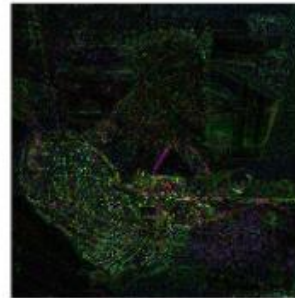
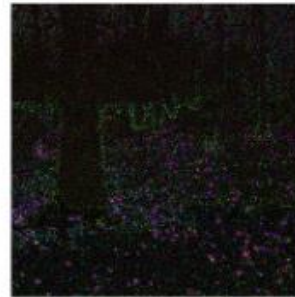
**Reconstructed  
Cover (Container)**



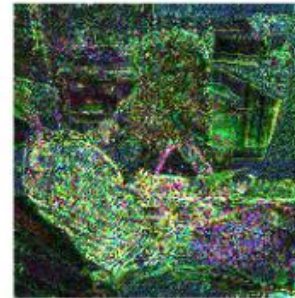
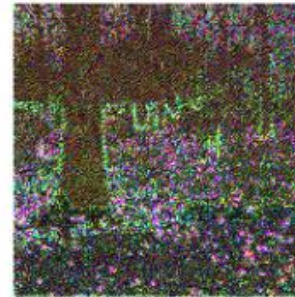
**Residual  
x5**



**x10**



**x20**




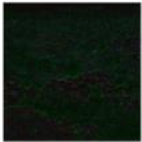




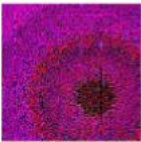

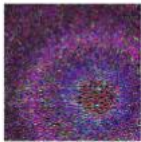

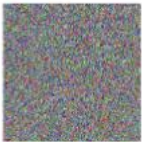

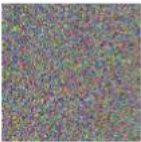

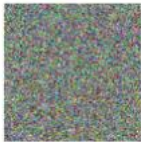


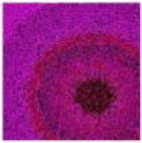

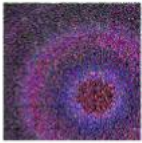
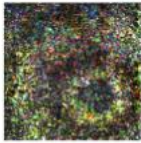
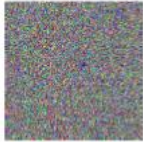



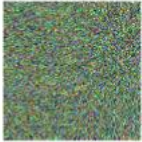
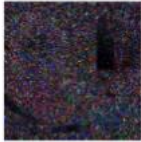


**Original  
Secret Image**





# Deep steganography

	Original Images Cover / Secret		Reconstructed Img Cover / Secret		Residual Error Cover / Secret		
Blank Cover Colorful Secret							Pixel Error: 6.0 / 8.8
Natural Cover Artificial Secret							8.3 / 28.8
Natural Cover Random Secret							7.1 / 45.1
Artificial Cover Natural Secret							28.9 / 29.6
Random Cover Natural Secret							39.4 / 17.2

# Deep steganography

