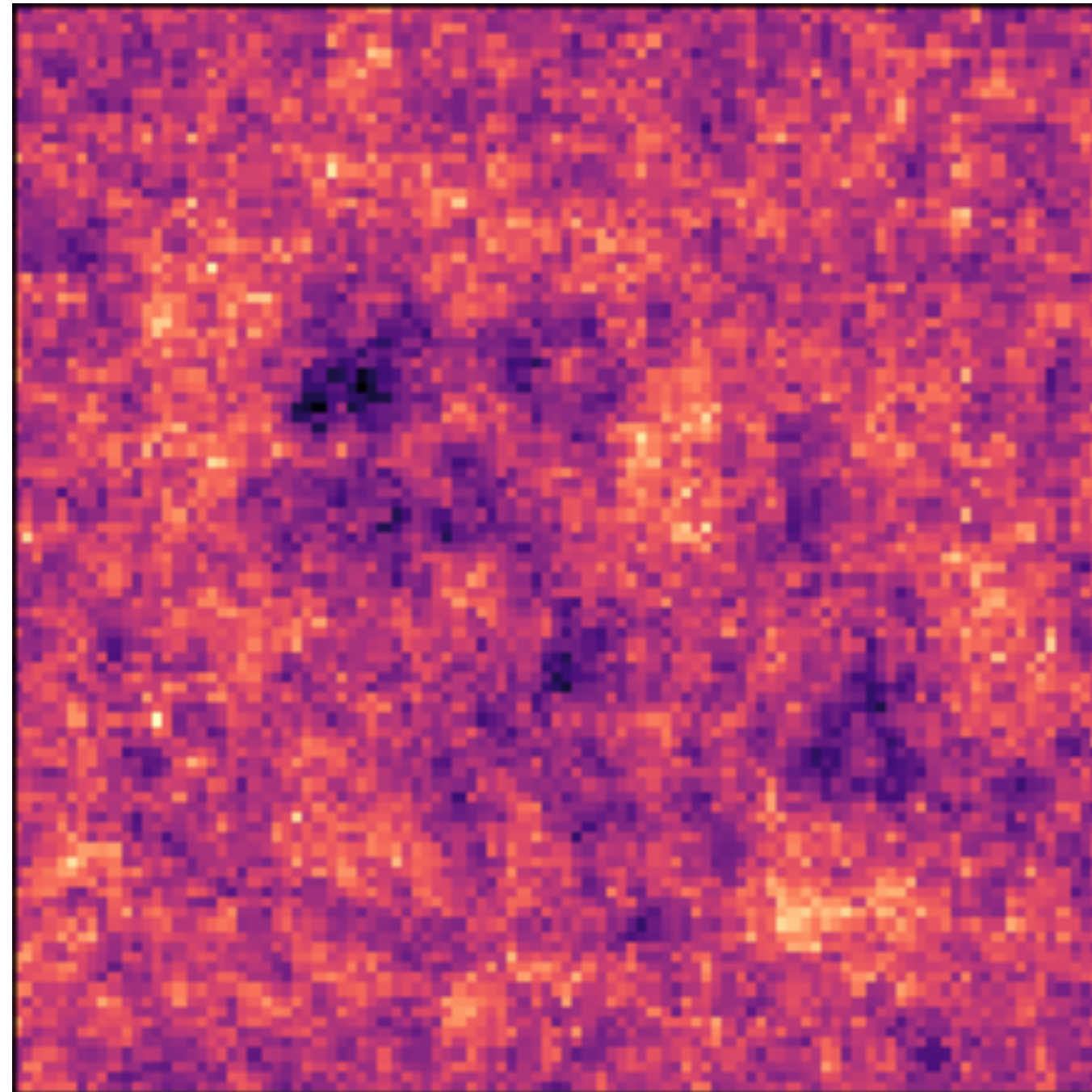


Noise2Self: Blind Denoising by Self-Supervision

Joshua Batson
Loïc Royer

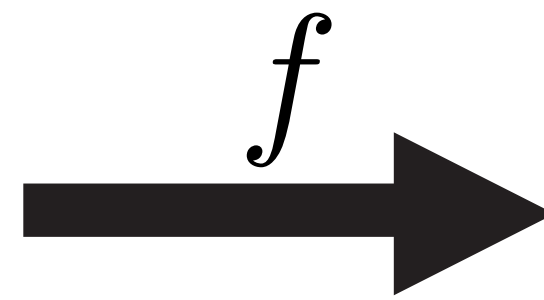
Noisy Data



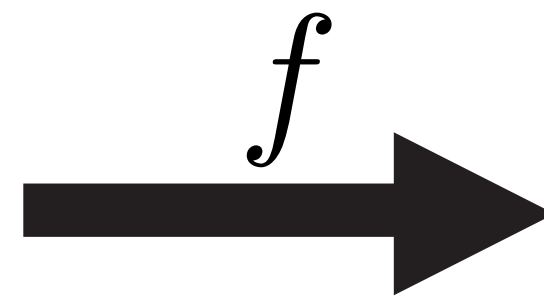
Supervision



Supervision



Supervision



$$\|f(x) - y\|^2$$

Self-Supervision?



Self-Supervision?



Self-Supervision?



$$\|f(x) - x\|^2$$

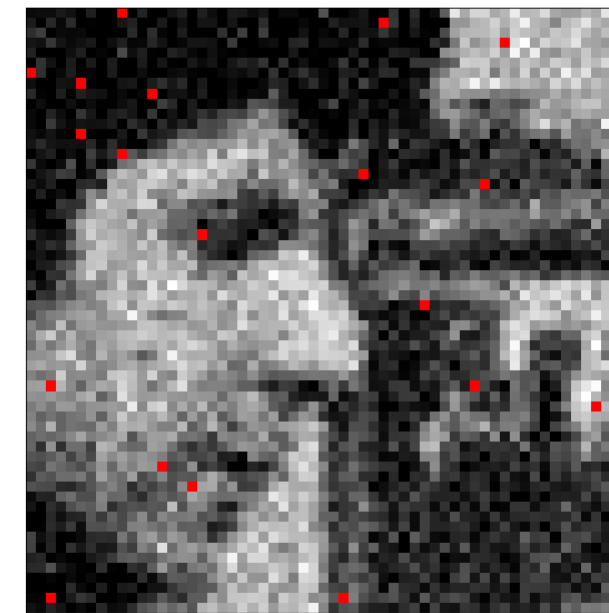
Self-Supervision?



$$\|f(x) - x\|^2$$

$$f^* = \text{Identity}$$

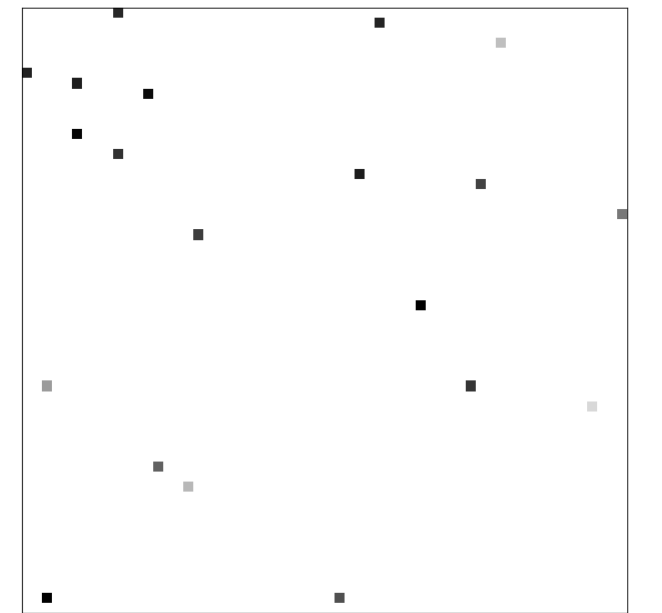
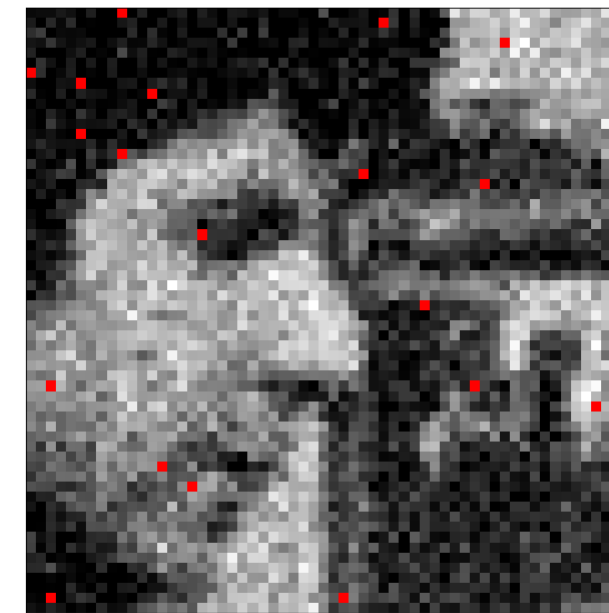
Self-Supervision?



$$\|f(x) - x\|^2$$

$$f^* = \text{Identity}$$

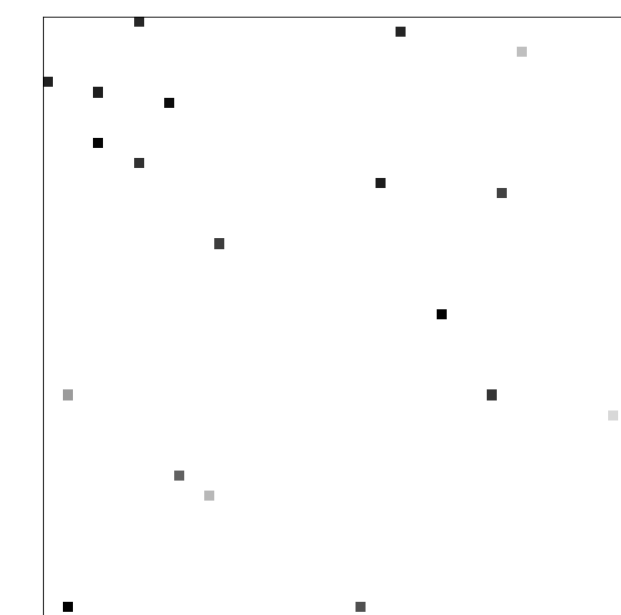
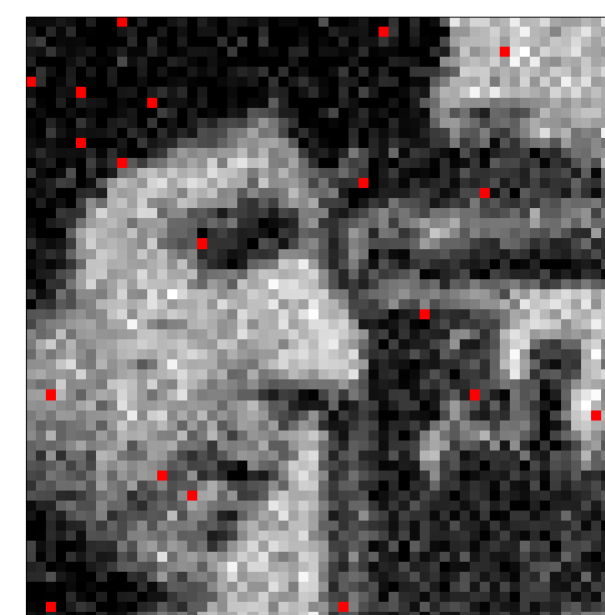
Self-Supervision?



$$\|f(x) - x\|^2$$

$$f^* = \text{Identity}$$

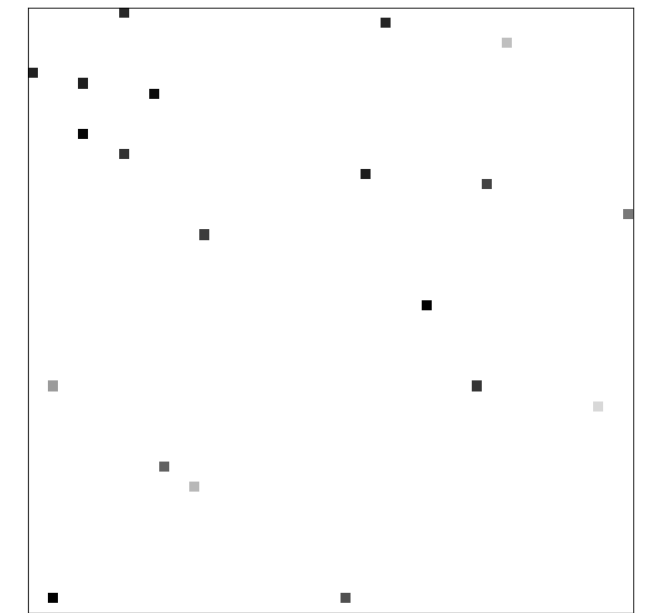
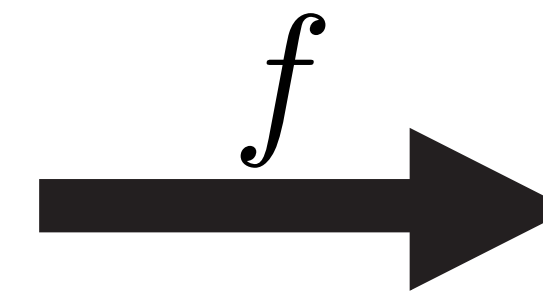
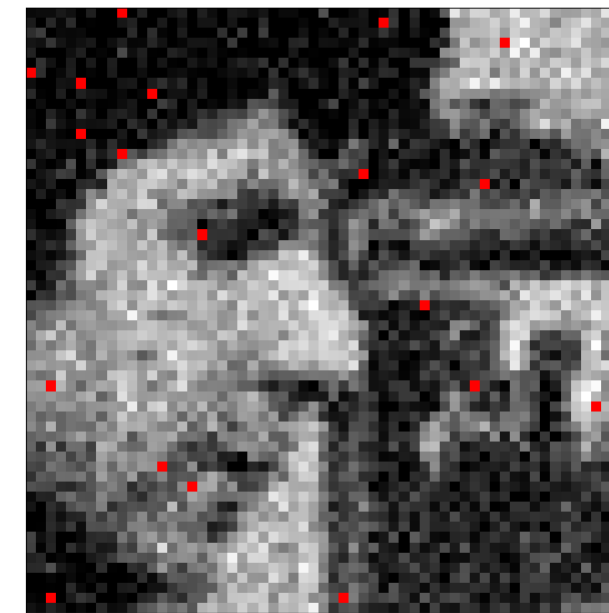
Self-Supervision?



$$\|f(x) - x\|^2$$

$$f^* = \text{Identity}$$

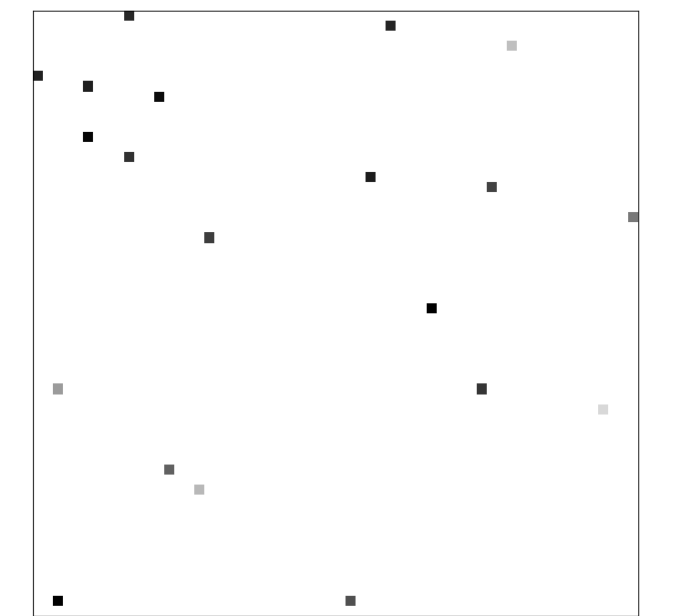
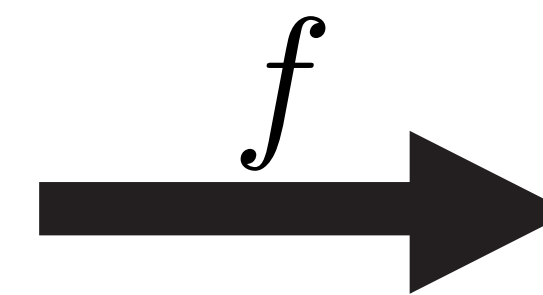
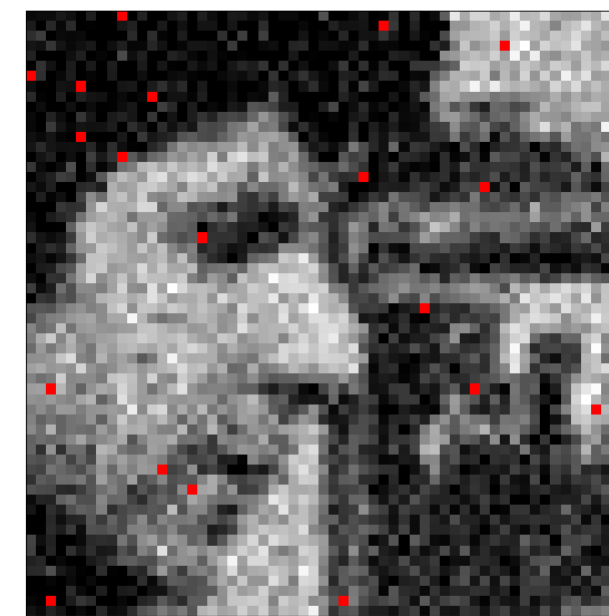
Self-Supervision?



$$\|f(x) - x\|^2$$

~~f^* Identity~~

Self-Supervision?

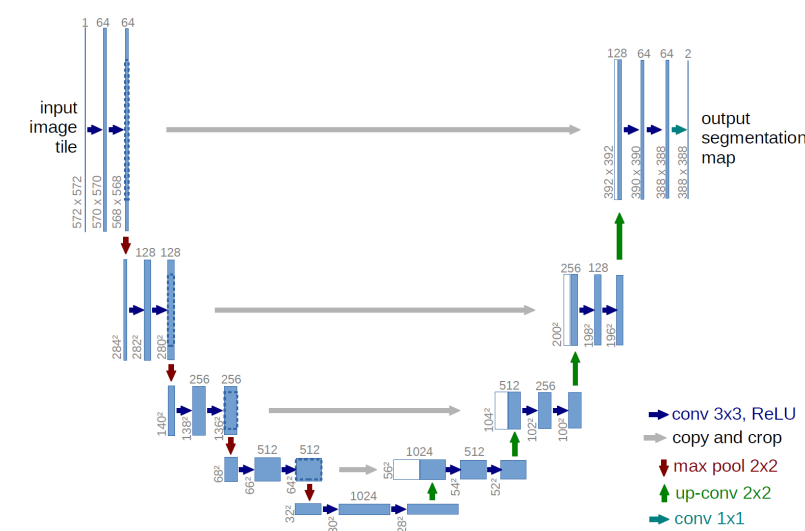


$$\|f(x) - x\|^2$$

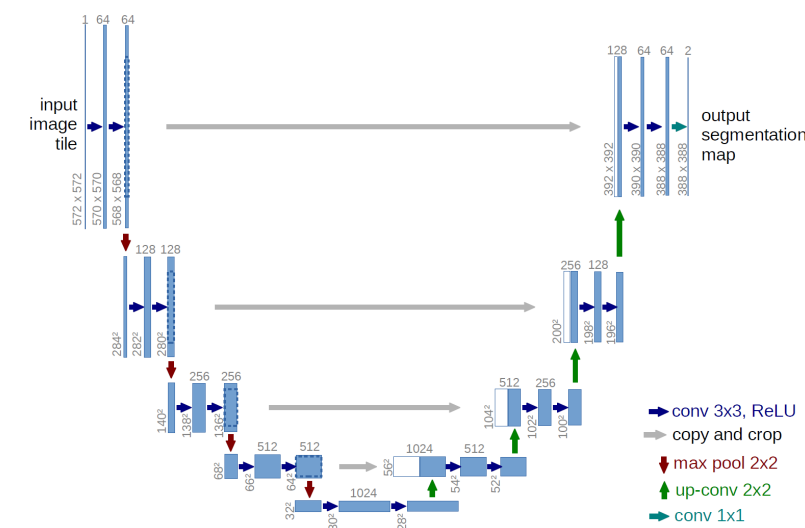
~~f^* Identity~~

$$f^* = \mathbb{E}[x_{-J} | x_J]$$

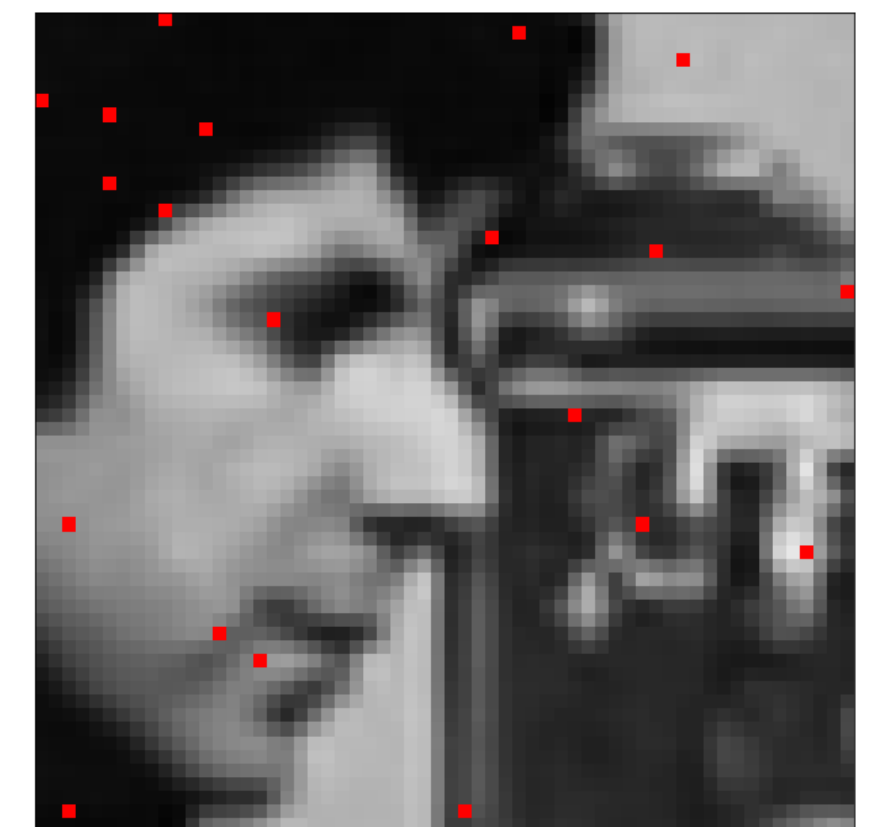
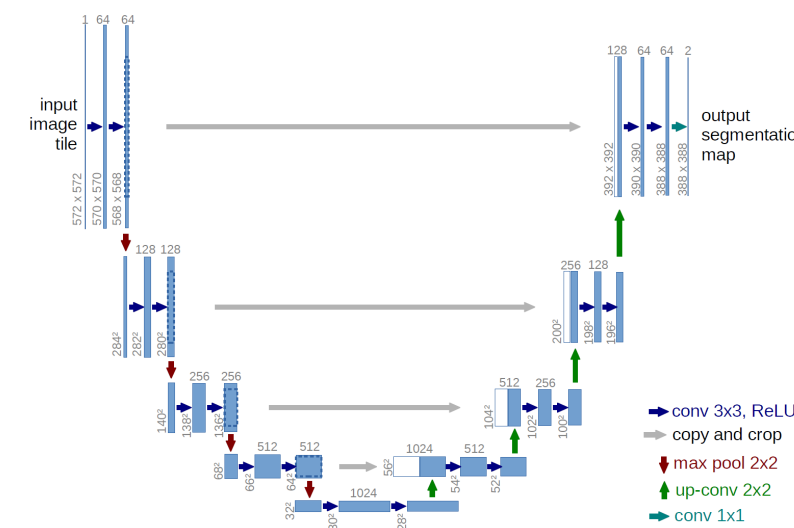
Single-Image Self-Supervised CNN Training



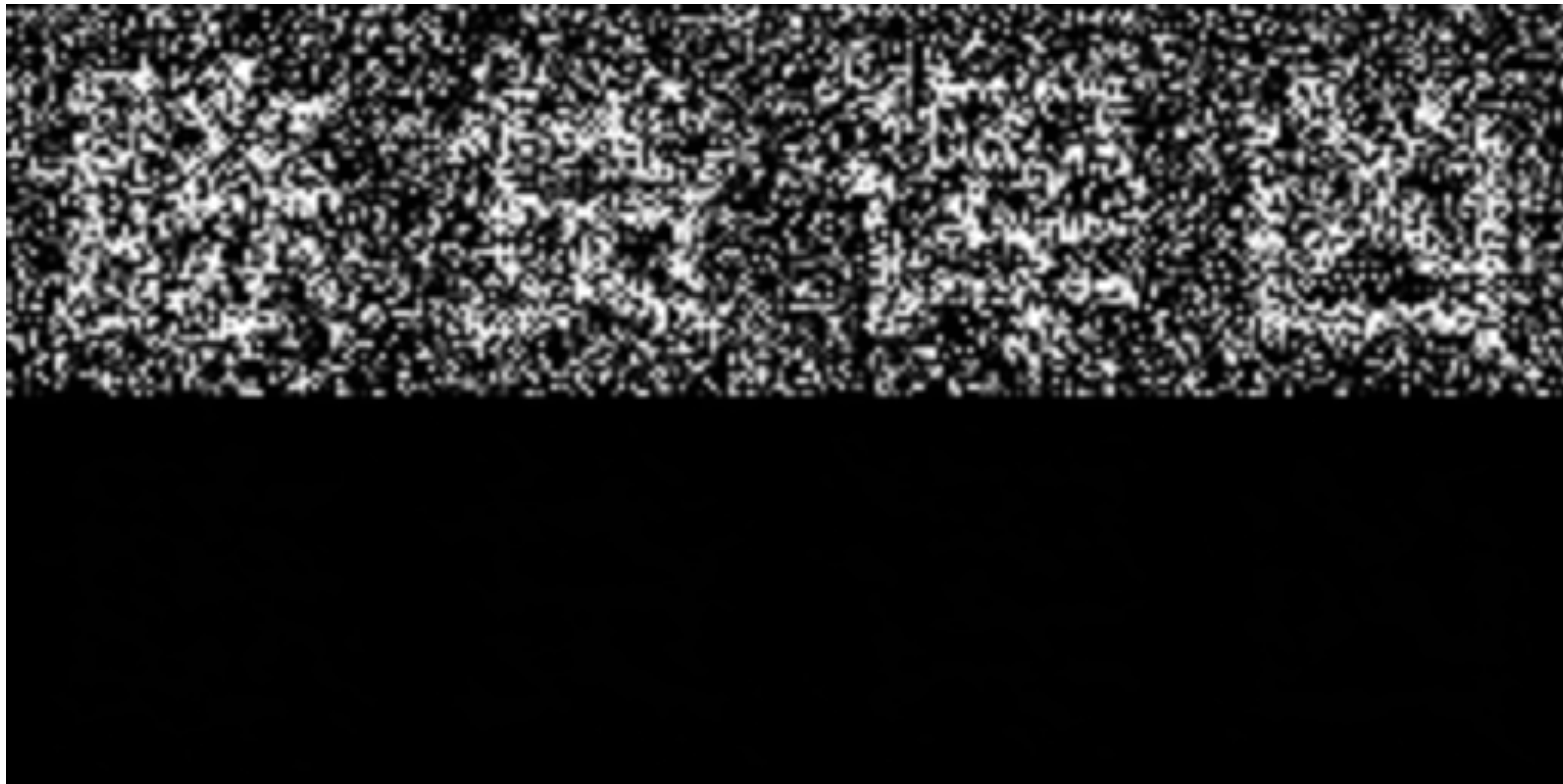
Single-Image Self-Supervised CNN Training



Single-Image Self-Supervised CNN Training



J-invariant Deep CNN



J-invariant Deep CNN



Plus...

Definitions

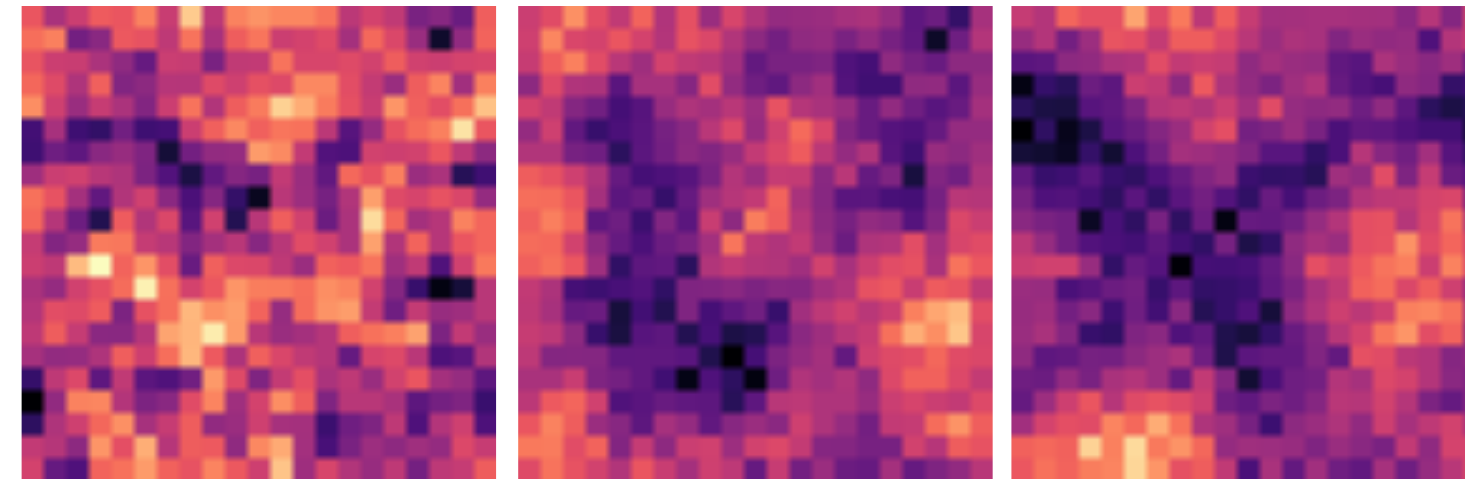
Definition. Let \mathcal{J} be a partition of the dimensions $\{1, \dots, m\}$ and let $J \in \mathcal{J}$. A function $f : \mathbb{R}^m \rightarrow \mathbb{R}^m$ is J -invariant if $f(x)_J$ does not depend on the value of x_J . It is \mathcal{J} -invariant if it is J -invariant for each $J \in \mathcal{J}$.

Theorems

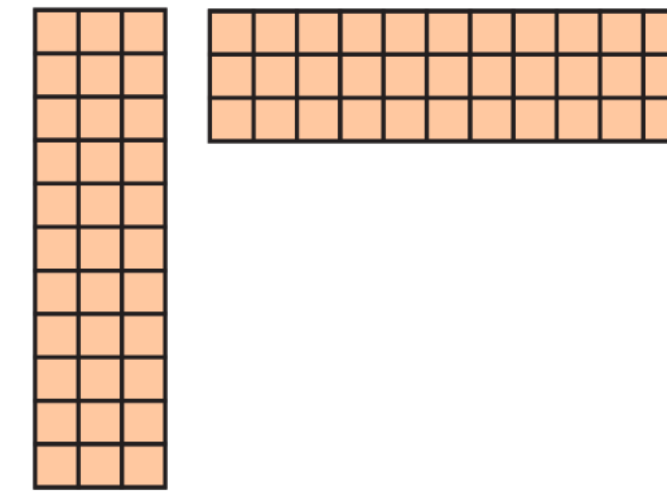
Proposition 3. Let x, y be random variables and let x^G and y^G be Gaussian random variables with the same covariance matrix. Let $f_{\mathcal{J}}^*$ and $f_{\mathcal{J}}^{*,G}$ be the corresponding optimal \mathcal{J} -invariant predictors. Then

$$\mathbb{E}\|y - f_{\mathcal{J}}^*(x)\|^2 \leq \mathbb{E}\|y - f_{\mathcal{J}}^{*,G}(x)\|^2.$$

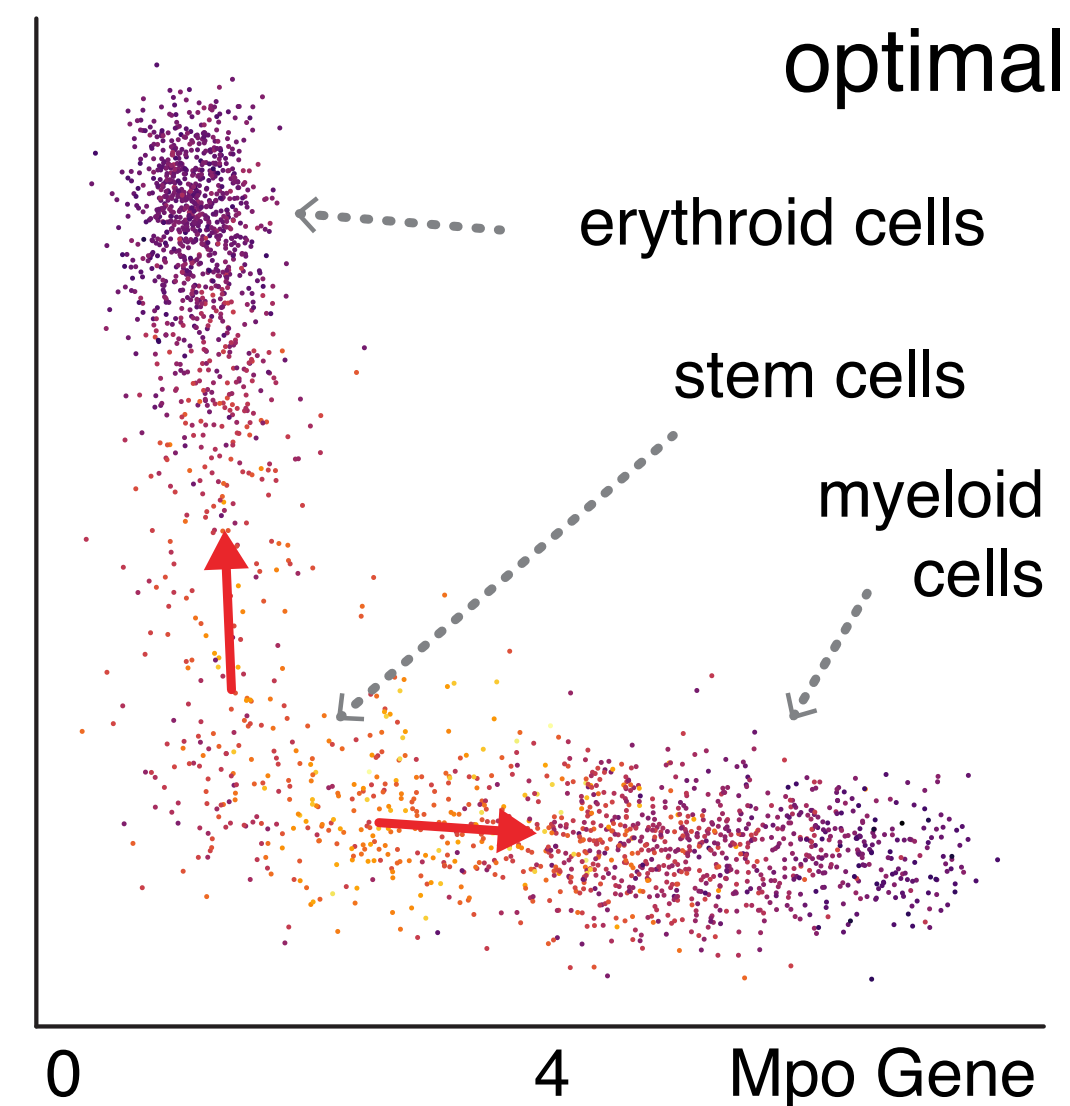
Gaussian Processes



Matrix Factorization

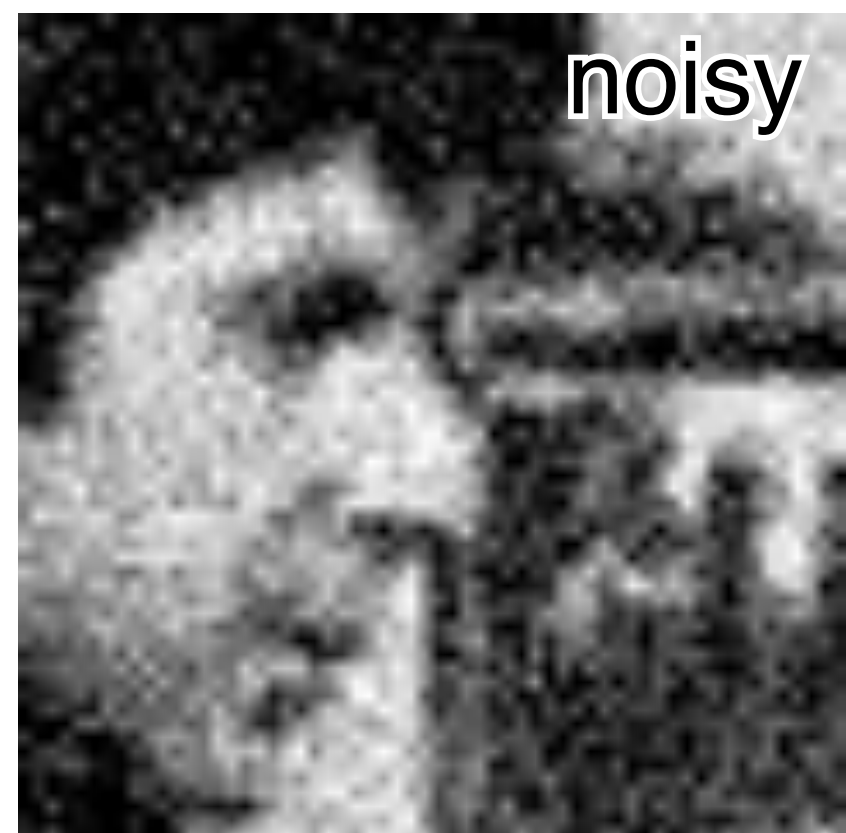


Single-Cell Sequencing

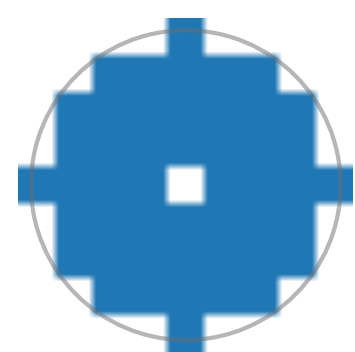


Code

```
for i, batch in enumerate(data_loader):
    noisy_images = batch
    input, mask = masker.mask(noisy_images, i)
    output = model(input)
    loss = loss_function(output*mask,
                          noisy_images*mask)
```

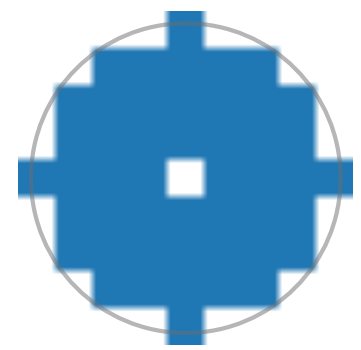


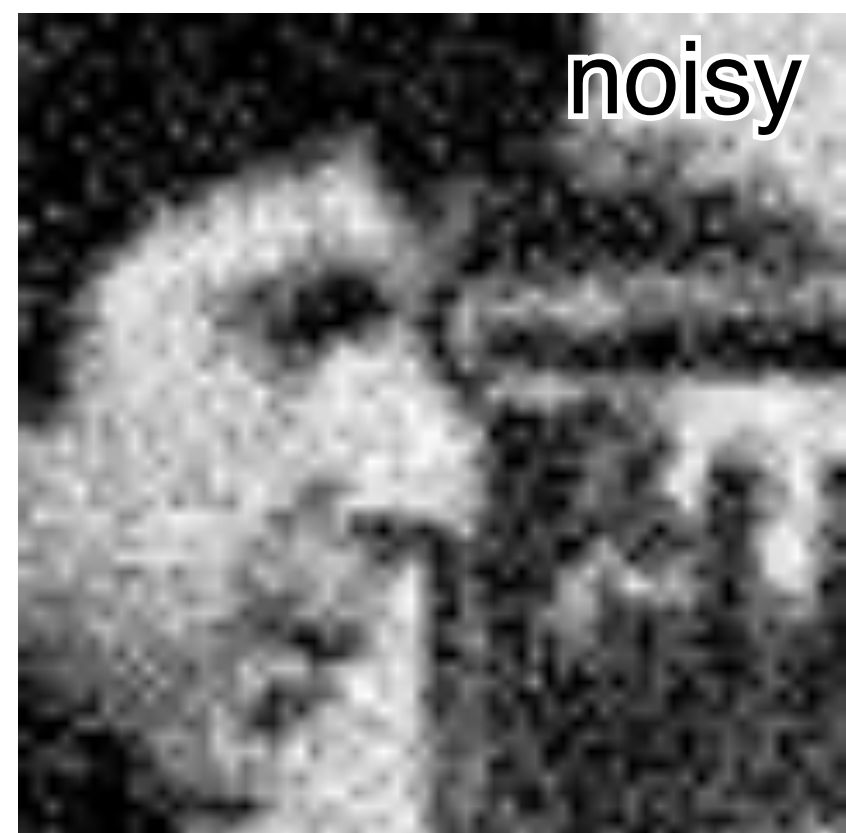
donut





donut





donut

