Machine Learning

Deep Learning with Auto-Encoding

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Acknowledge: The slides are partially referred to the online materials by Prof. Hung-yi Lee from NTU, Auto-encoder session, and also other online materials Note that some figures and contents used in the presentation are from public available resources, the original authors own the copyright of the figures and content

Unsupervised Learning

"We expect unsupervised learning to become far more important in the longer term. Human and animal learning is largely unsupervised: we discover the structure of the world by observing it, not by being told the name of every object."

- LeCun, Bengio, Hinton, Nature 2015

As I've said in previous statements: most of human and animal learning is unsupervised learning. If intelligence was a cake, unsupervised learning would be the cake, supervised learning would be the icing on the cake, and reinforcement learning would be the cherry on the cake. We know how to make the icing and the cherry, but we don't know how to make the cake.

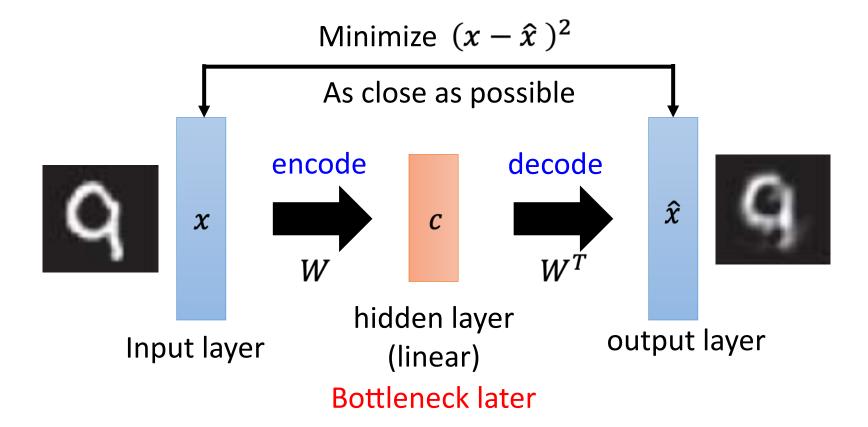
- Yann LeCun, March 14, 2016 (Facebook)

Unsupervised Learning

What we have learnt:

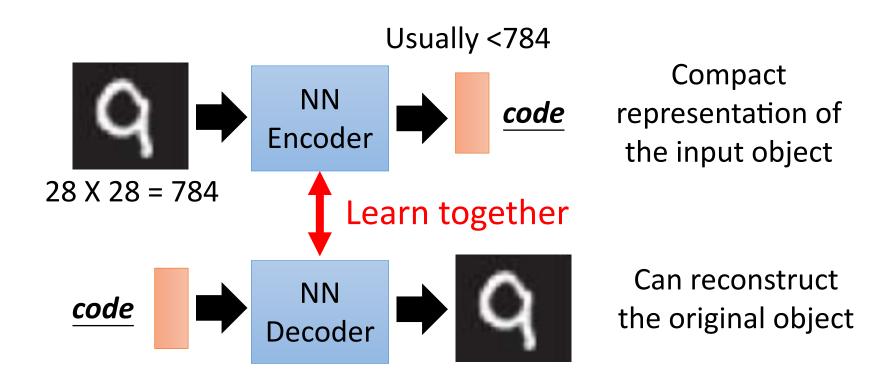
Manifold Learning: PCA, LLE, MDS, etc.

Recap: PCA



Output of the hidden layer is the code

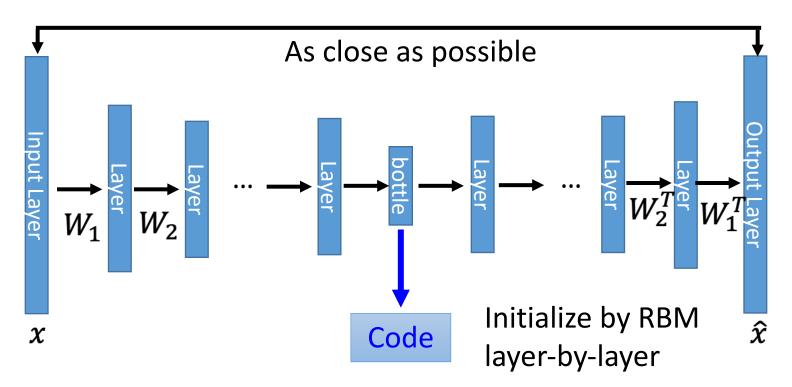
Auto-encoder



Deep Auto-encoder

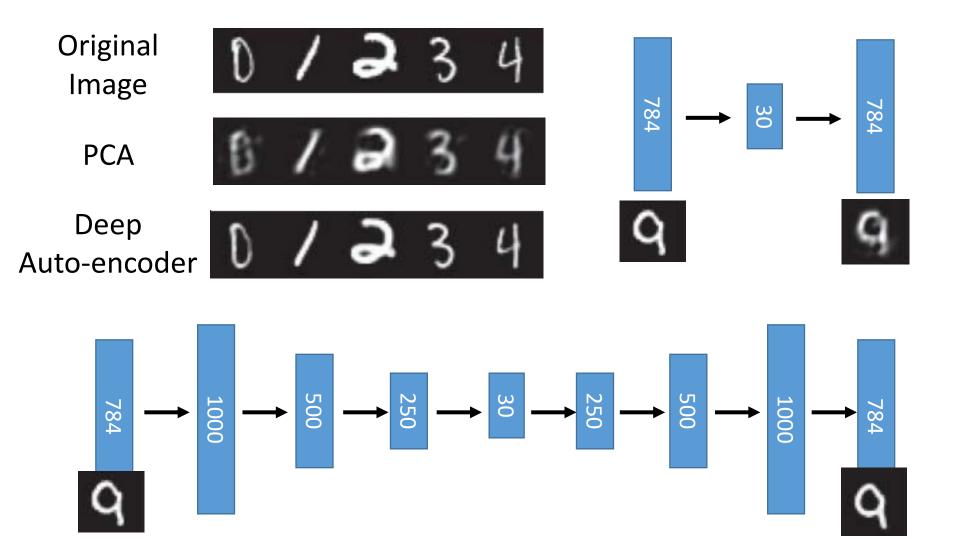
Symmetric is not necessary.

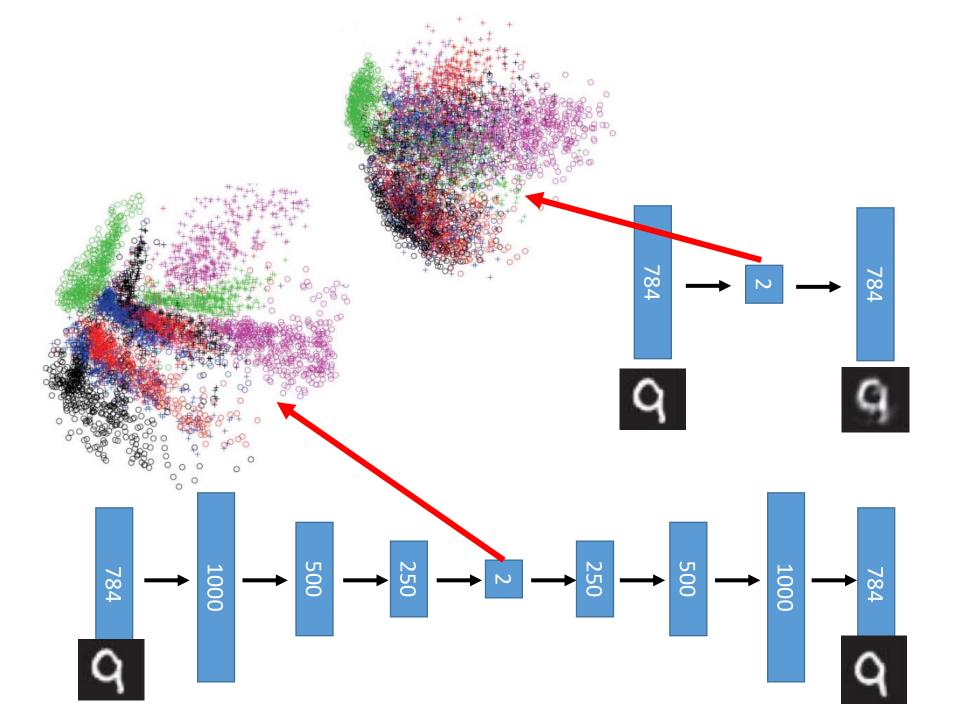
Of course, the auto-encoder can be deep



Reference: Hinton, Geoffrey E., and Ruslan R. Salakhutdinov. "Reducing the dimensionality of data with neural networks." *Science* 313.5786 (2006): 504-507

Deep Auto-encoder



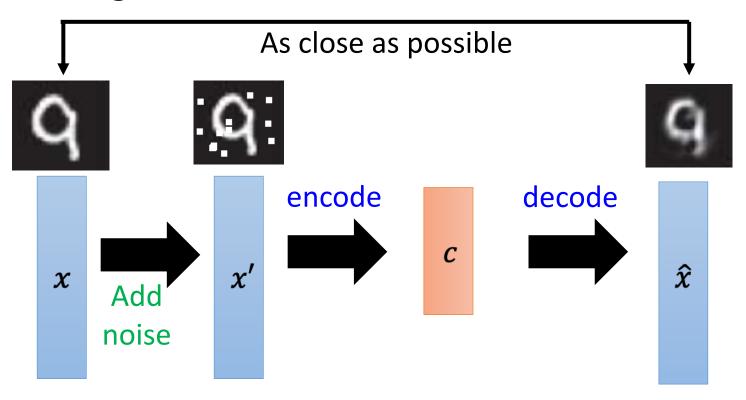


More: Contractive auto-encoder

Auto-encoder

Ref: Rifai, Salah, et al. "Contractive auto-encoders: Explicit invariance during feature extraction." *Proceedings of the 28th International Conference on Machine Learning (ICML-11)*. 2011.

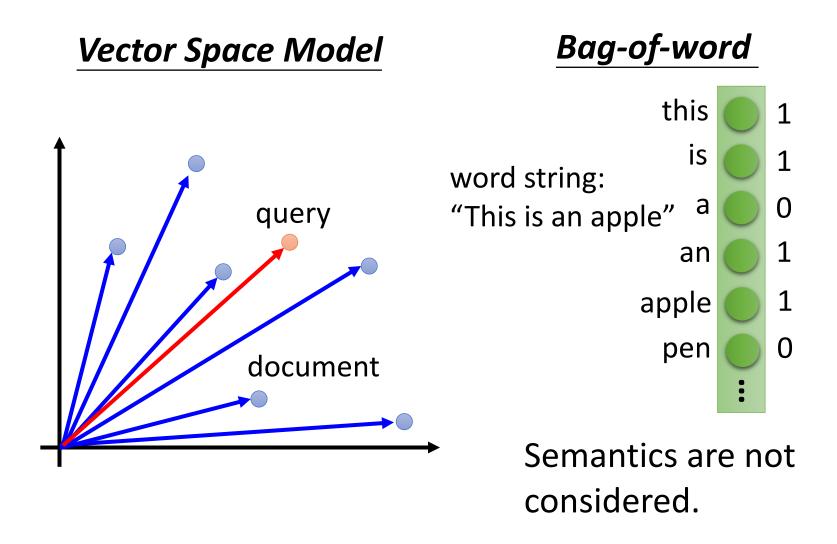
De-noising auto-encoder



Vincent, Pascal, et al. "Extracting and composing robust features with denoising autoencoders." *ICML*, 2008.

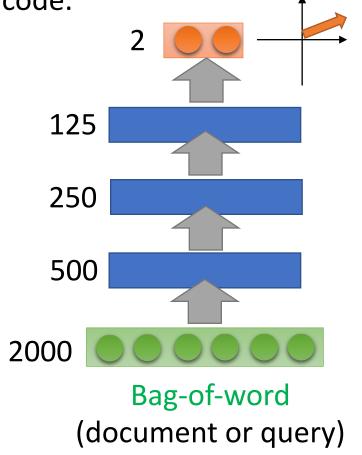
Deep Auto-encoder - Example NN Encoder **PCA** Reduced 32dim Pixel -> tSNE

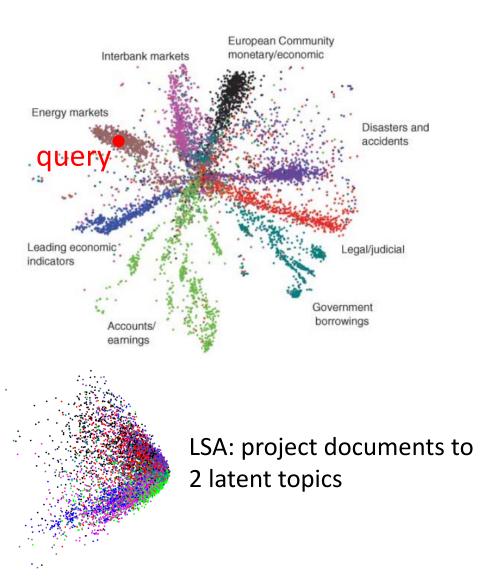
Auto-encoder – Text Retrieval



Auto-encoder – Text Retrieval

The documents talking about the same thing will have close code.





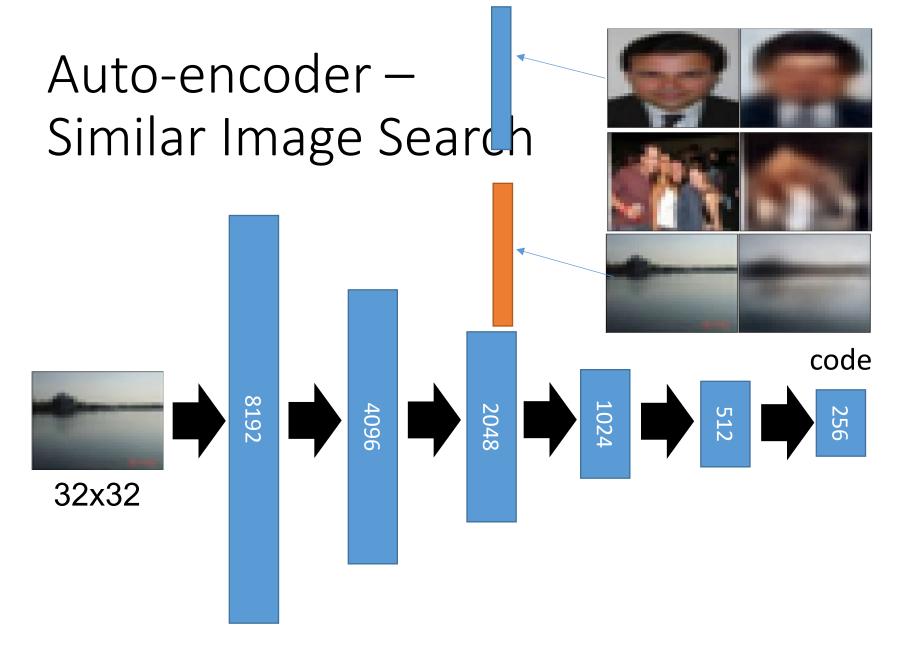
Auto-encoder – Similar Image Search

Retrieved using Euclidean distance in pixel intensity space



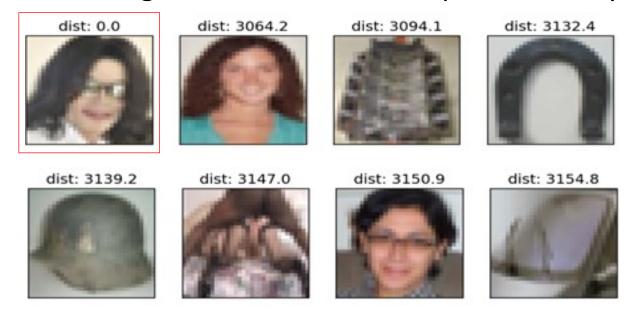
(Images from Hinton's slides on Coursera)

Reference: Krizhevsky, Alex, and Geoffrey E. Hinton. "Using very deep autoencoders for content-based image retrieval." *ESANN*. 2011.



(crawl millions of images from the Internet)

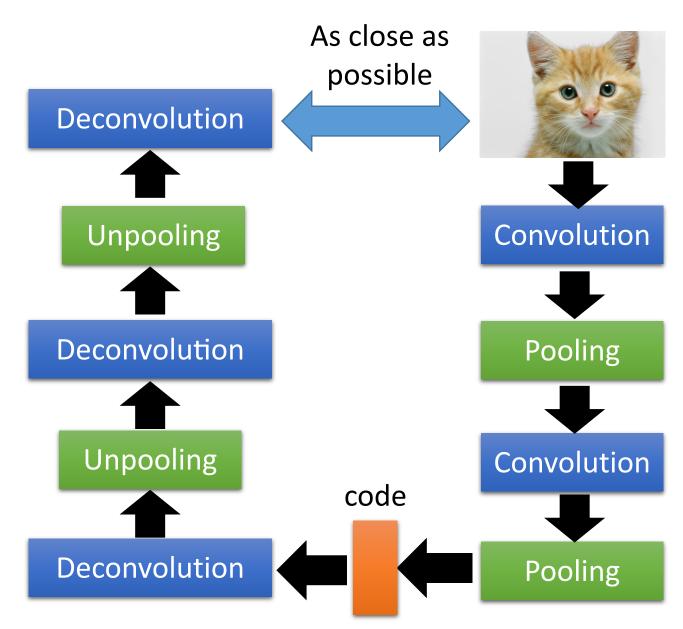
Retrieved using Euclidean distance in pixel intensity space



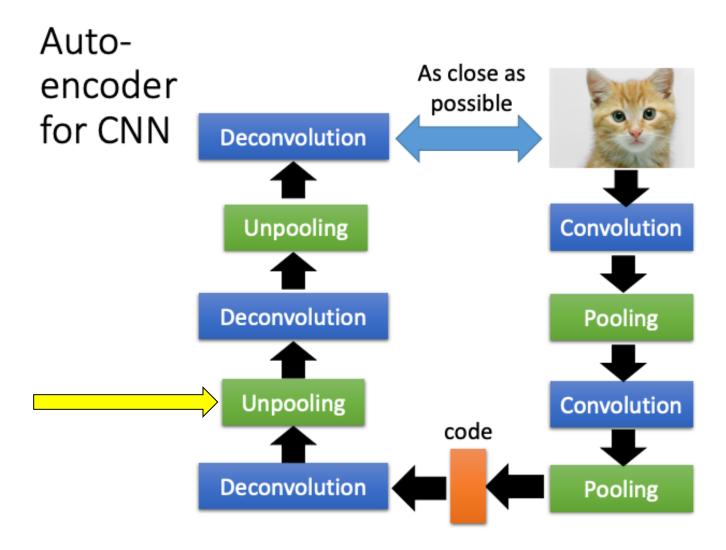
retrieved using 256 codes



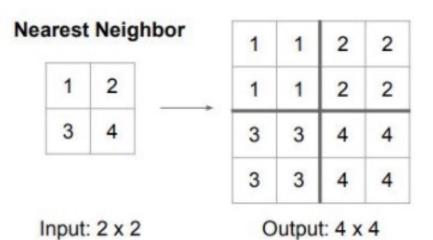
Autoencoder for CNN

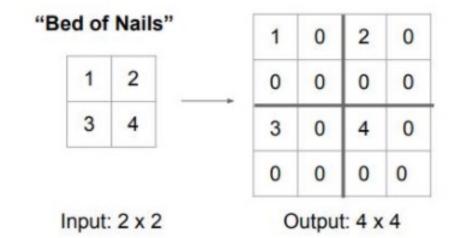


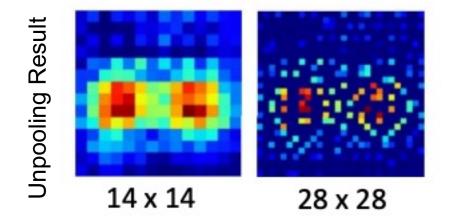
CNN For Auto-Encoder



Unpooling

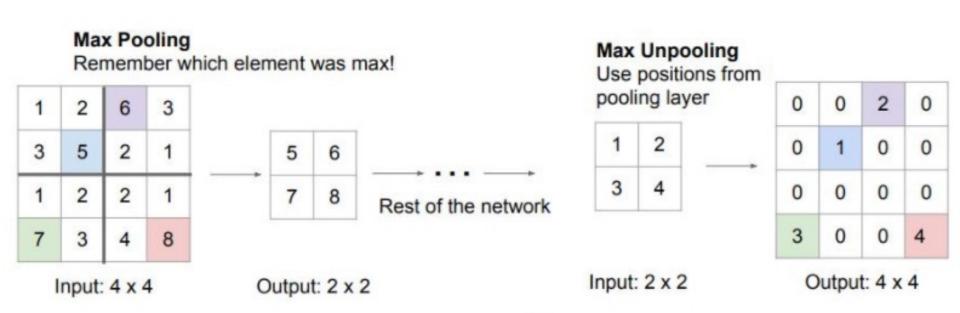




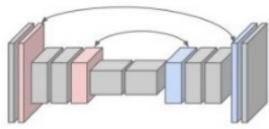


Source: https://jinzequn.github.io/2018/01/28/deconv-and-unpool/

Max Unpooling

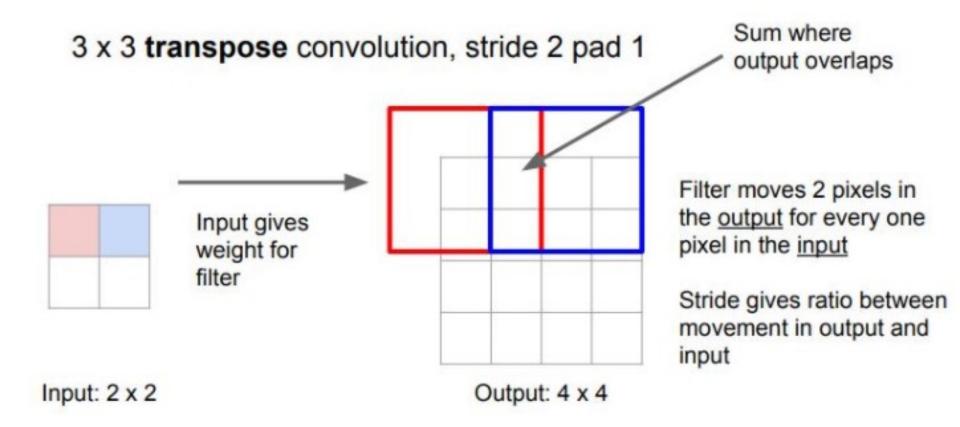


Corresponding pairs of downsampling and upsampling layers

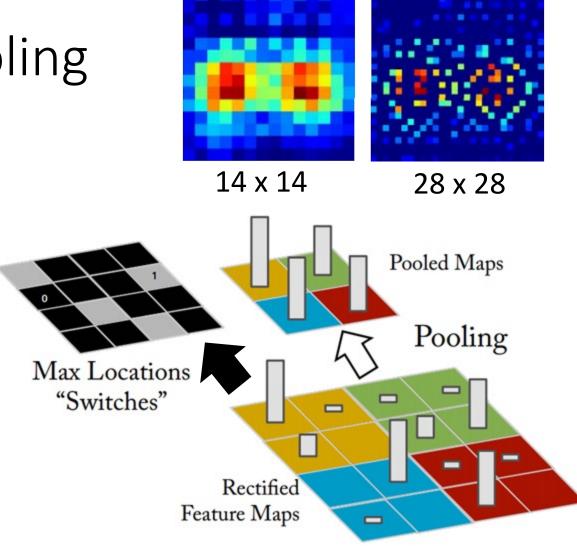


Source: https://jinzequn.github.io/2018/01/28/deconv-and-unpool/

Transpose Convolution



CNN -Unpooling

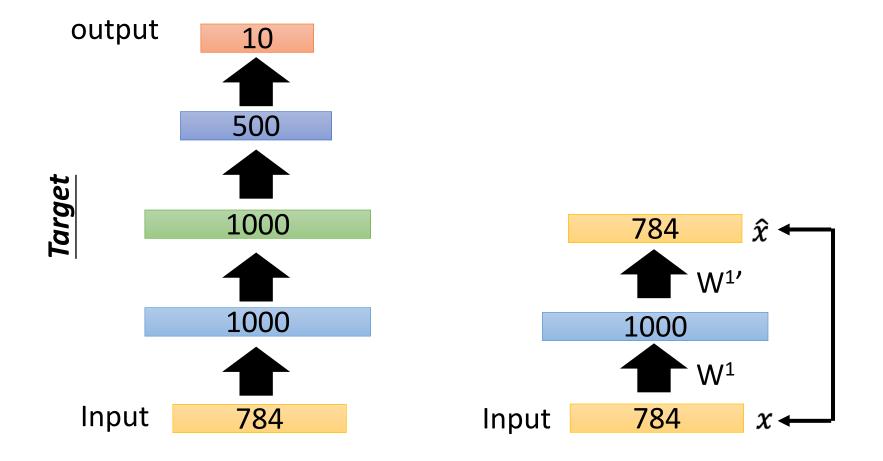


Alternative: simply repeat the values

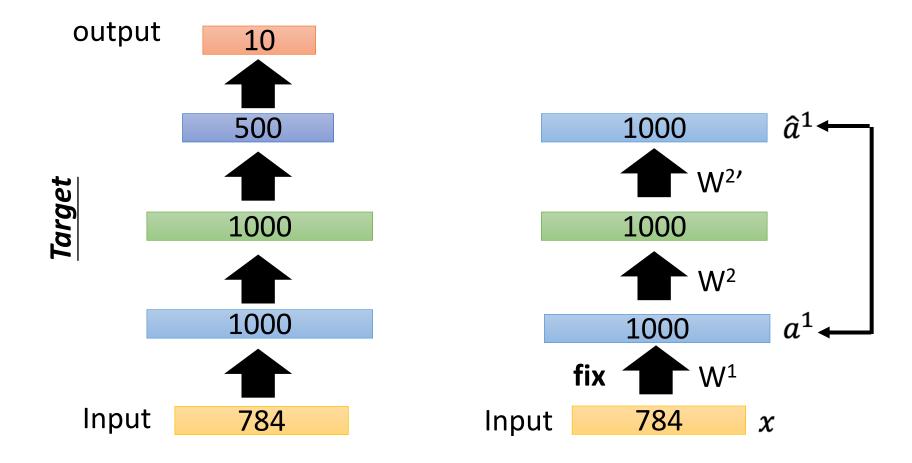
Source of image:

https://leonardoaraujosantos.gitbooks.io/artificial-inteligence/content/image_segmentation.html

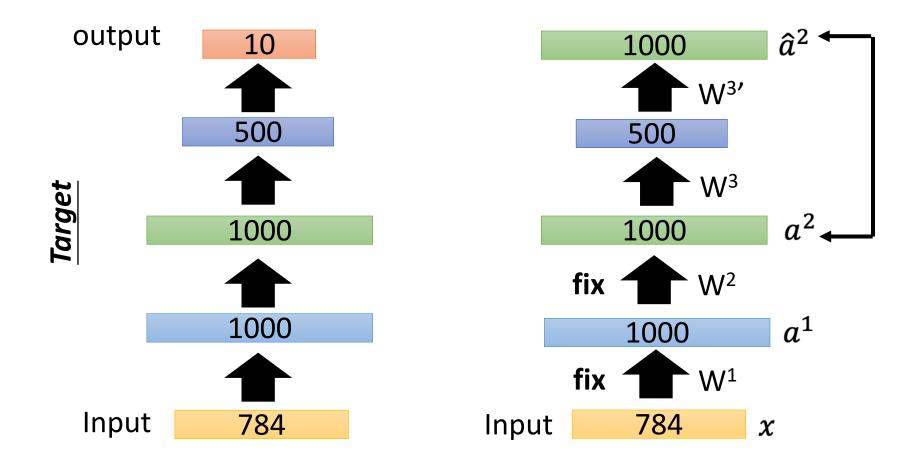
Greedy Layer-wise Pre-training again



Greedy Layer-wise Pre-training again

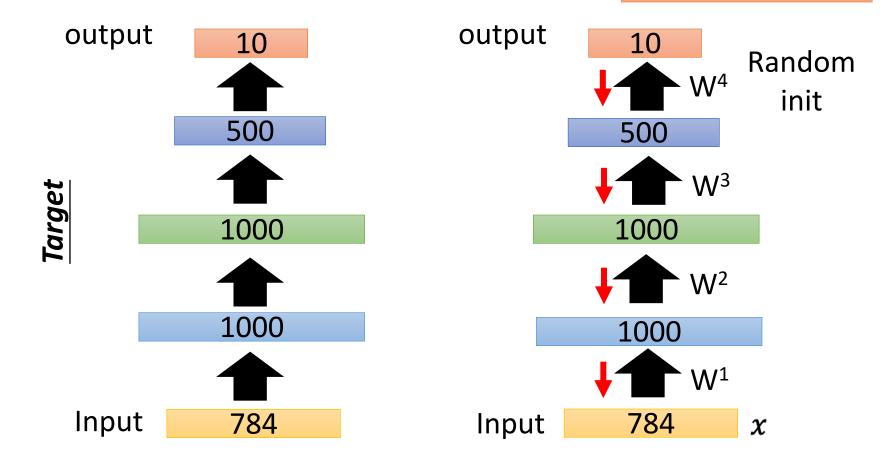


Greedy Layer-wise Pre-training again

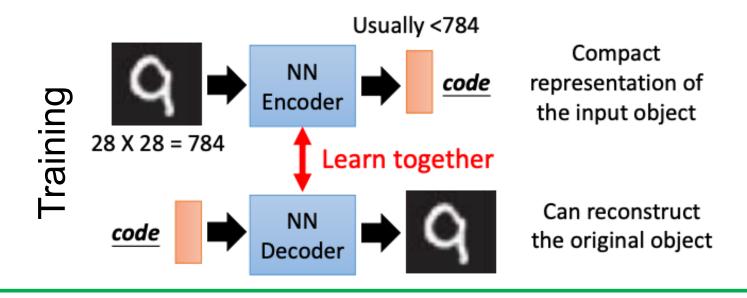


• Greedy Layer-wise Pre-training again

Find-tune by backpropagation



Can we use Decoder to Generate Something?



Senerator

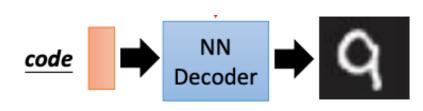


Image Generator

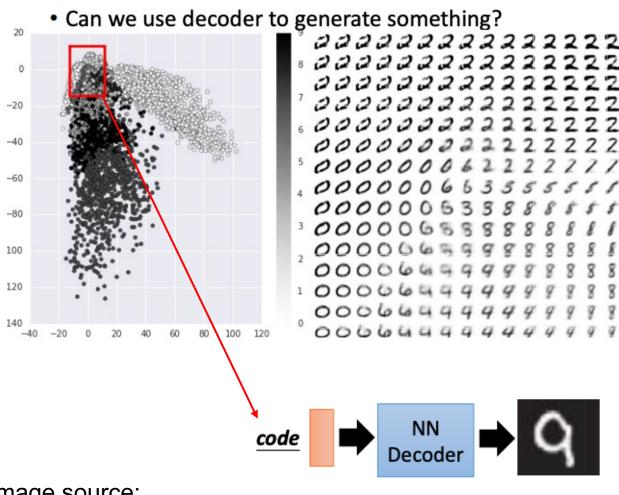


Image source: