Unsupervised Learning:
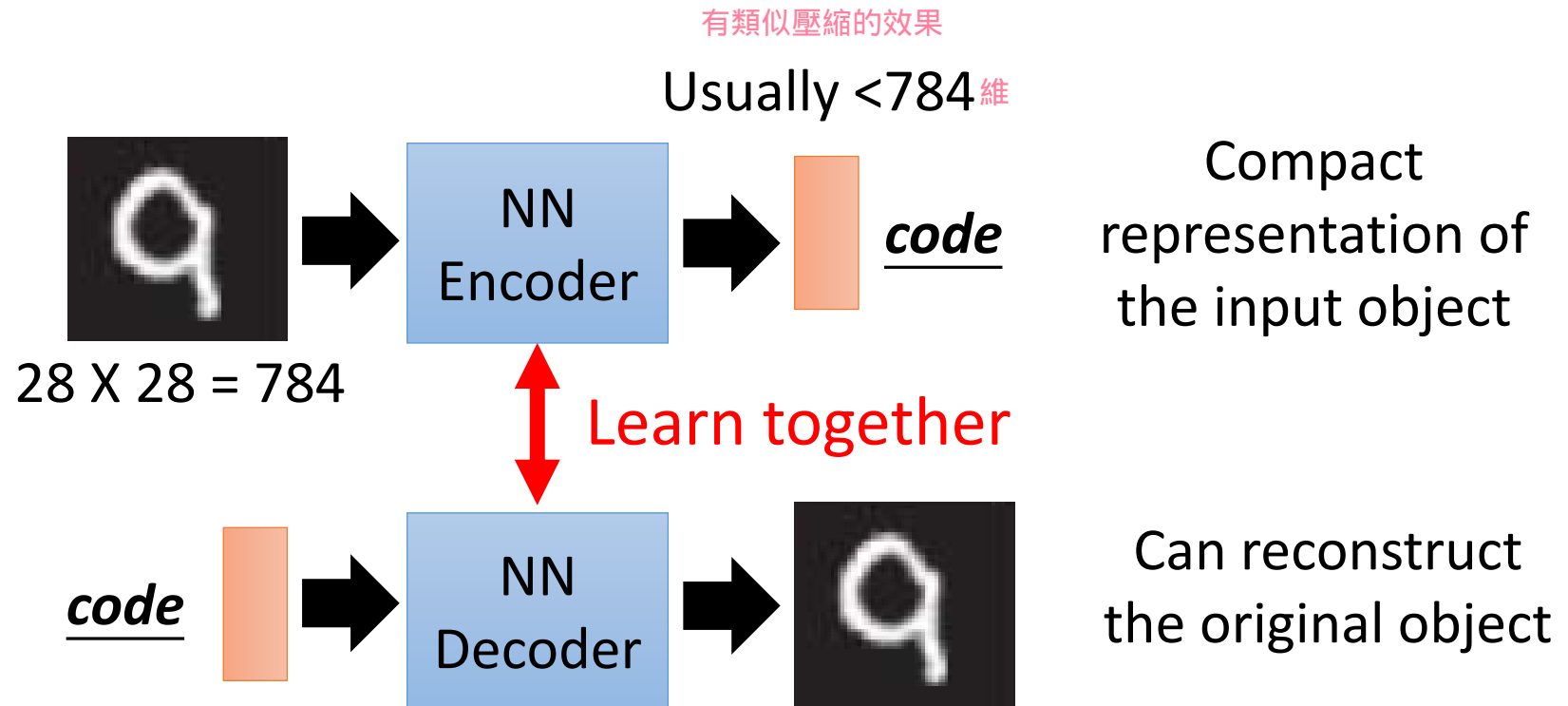## Deep Auto-encoder

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."
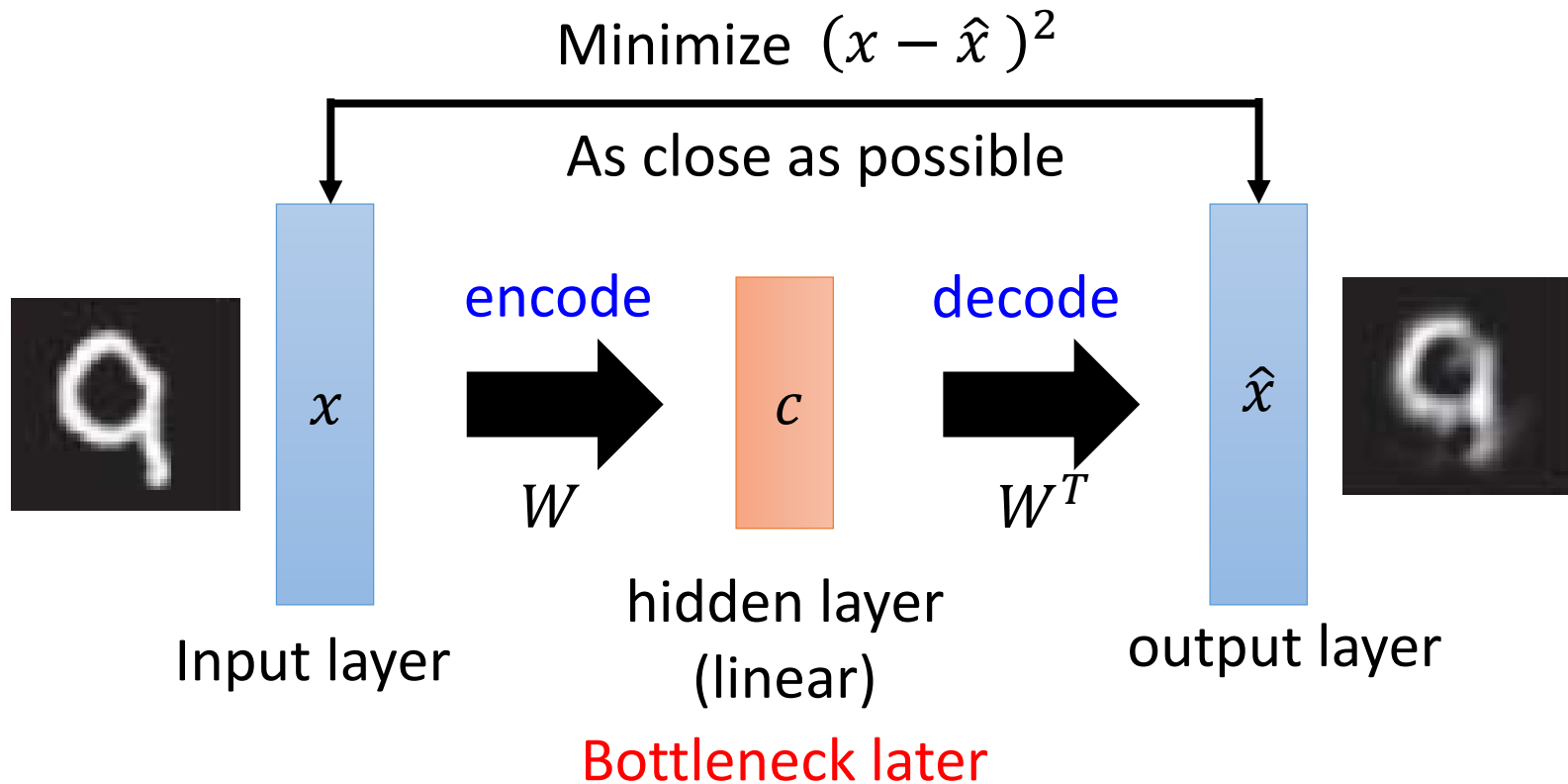– LeCun, Bengio, Hinton, Nature 2015

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.
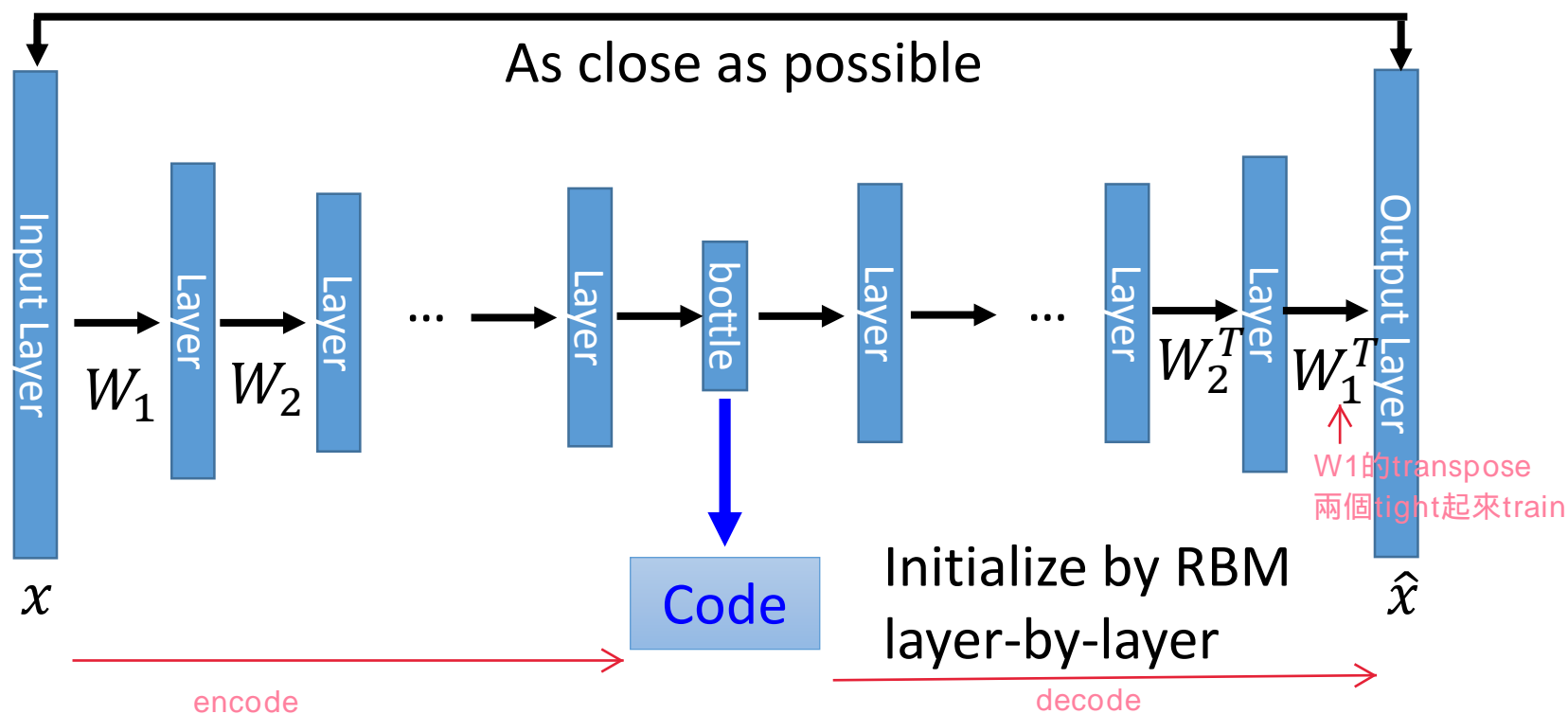- Yann LeCun, March 14, 2016 (Facebook)

Auto-encoder

Usually <784

28 X 28 = 784

NN Encoder

code

Compact representation of the input object

Learn together

code

NN Decoder

Can reconstruct the original object

Recap: PCA

Minimize $(x - \hat{x})^2$

As close as possible

encode

decode

x

c

$\hat{x}$

W

W^T

Input layer

hidden layer
(linear)

output layer

Bottleneck later

Output of the hidden layer is the code

Deep Auto-encoder

Symmetric is not necessary.

train   weight

• Of course, the auto-encoder can be deep

As close as possible

Input Layer → Layer → Layer → ... → Layer → bottle → Layer → ... → Layer → Layer → Output Layer

W_1 W_2 W_2^T W_1^T

W1   transpose
tight   train

x

Code

Initialize by RBM layer-by-layer

$\hat{x}$

encode

decode

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

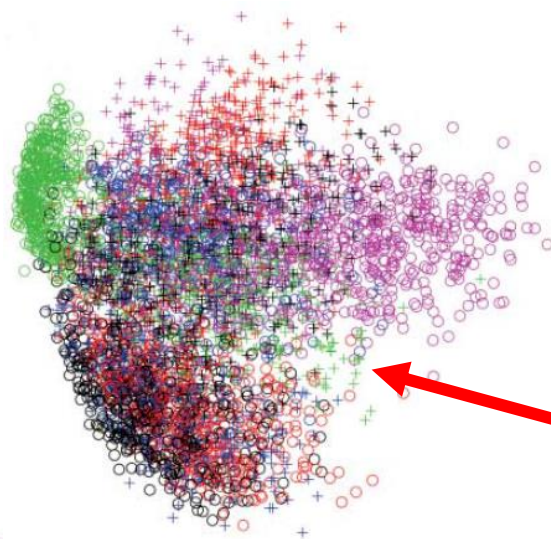# Deep Auto-encoder

Original
Image

PCA

Deep
Auto-encoder

PCA

auto-encoder
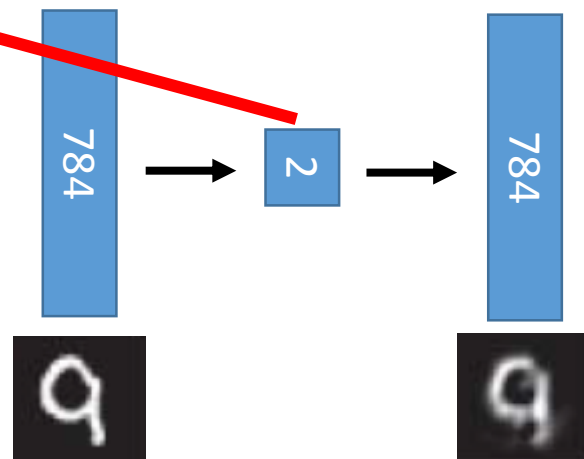
PCA

auto-encoder

784 → 2 → 784

784 → 1000 → 500 → 250 → 2 → 250 → 500 → 1000 → 784

Auto-encoder
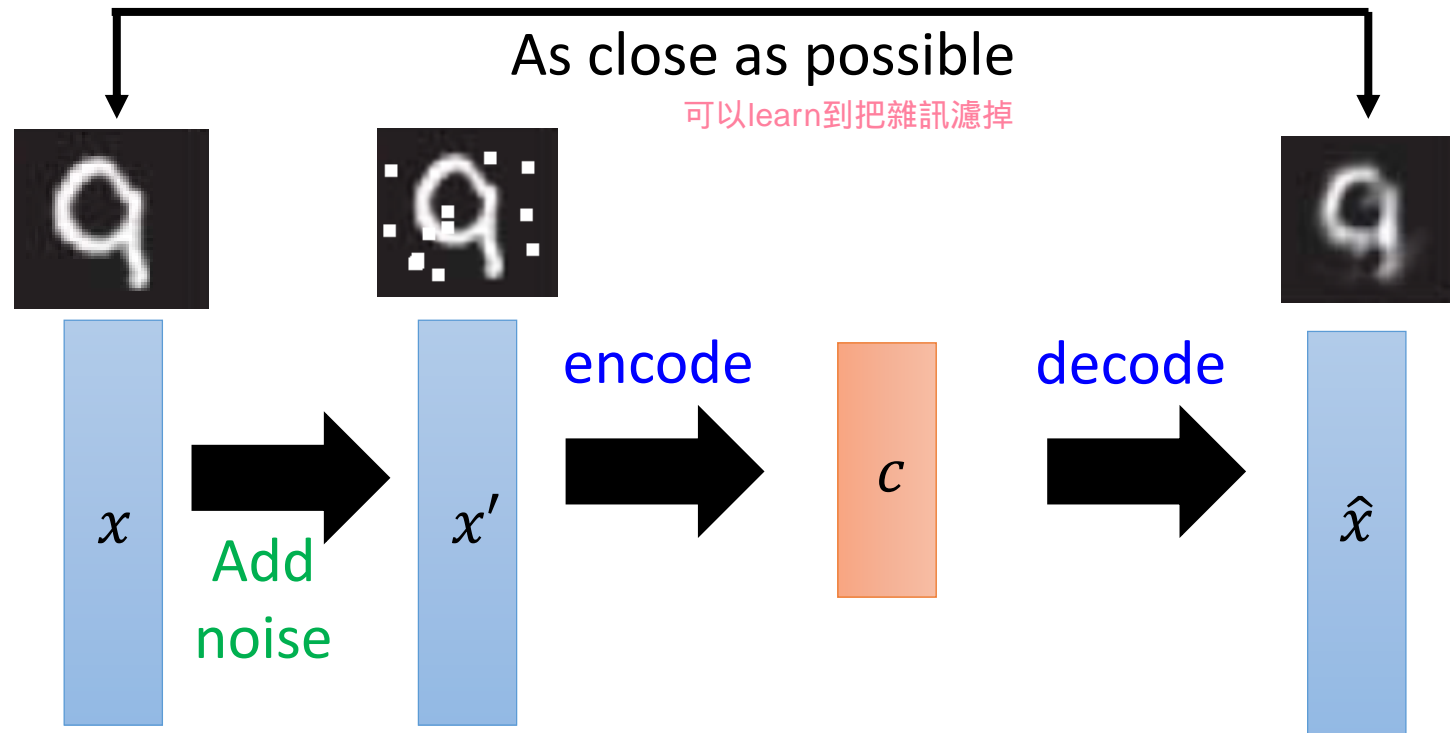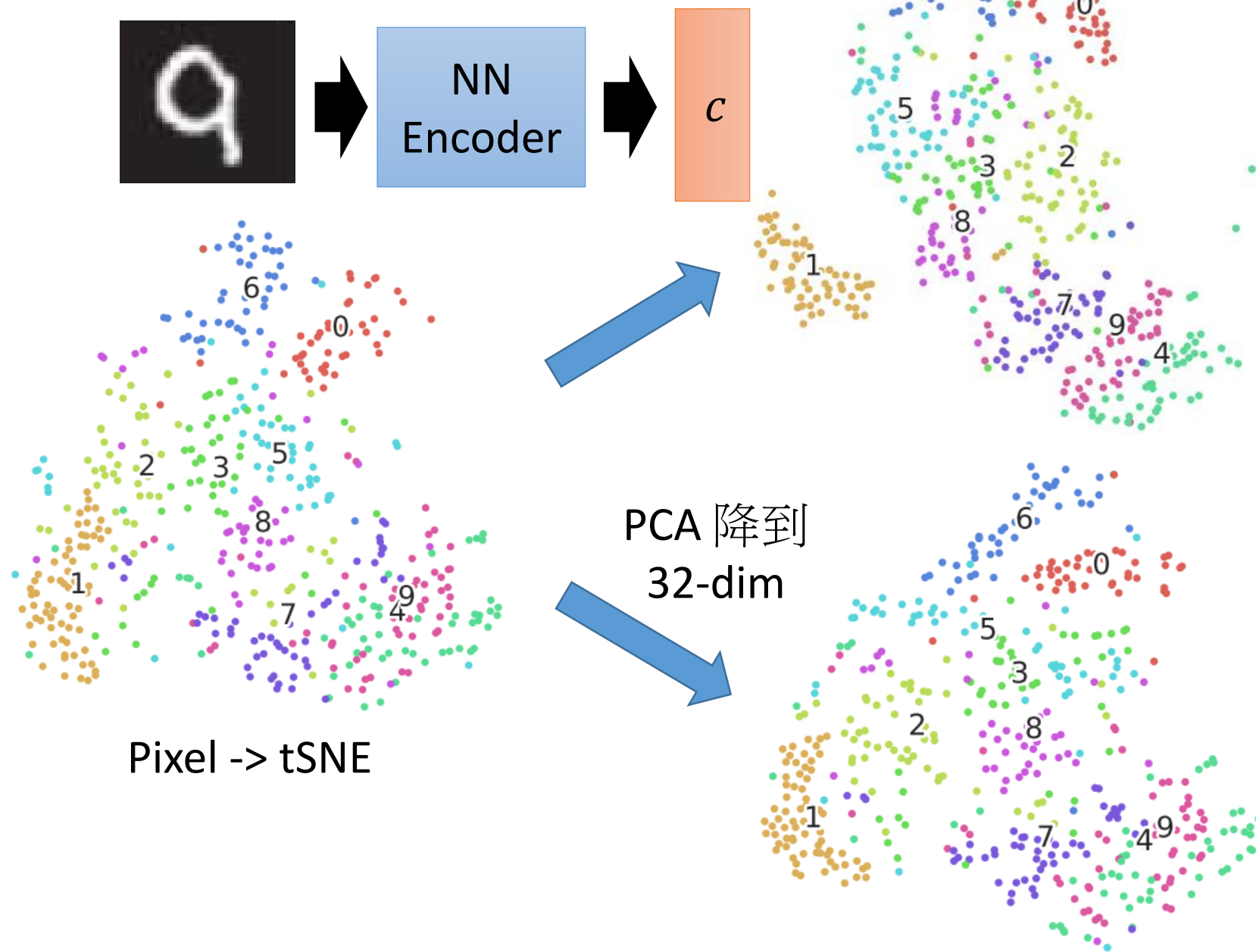
More: Contractive auto-encoder

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

As close as possible

learn

x Add noise x' encode c decode $\hat{x}$

Vincent, Pascal, et al. "Extracting and composing robust features with denoising autoencoders." *ICML,* 2008.
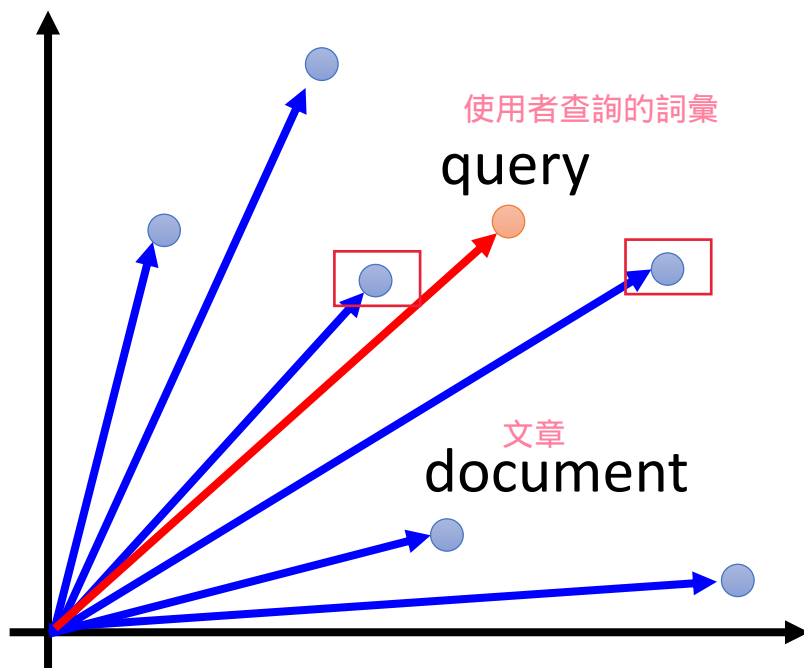
Deep Auto-encoder - Example

NN Encoder → c

PCA 降到 32-dim

Pixel -> tSNE

Auto-encoder – Text Retrieval

vector

Vector Space Model

vector

Bag-of-word

query

document

cosine similarity   (   )

word string:
"This is an apple"

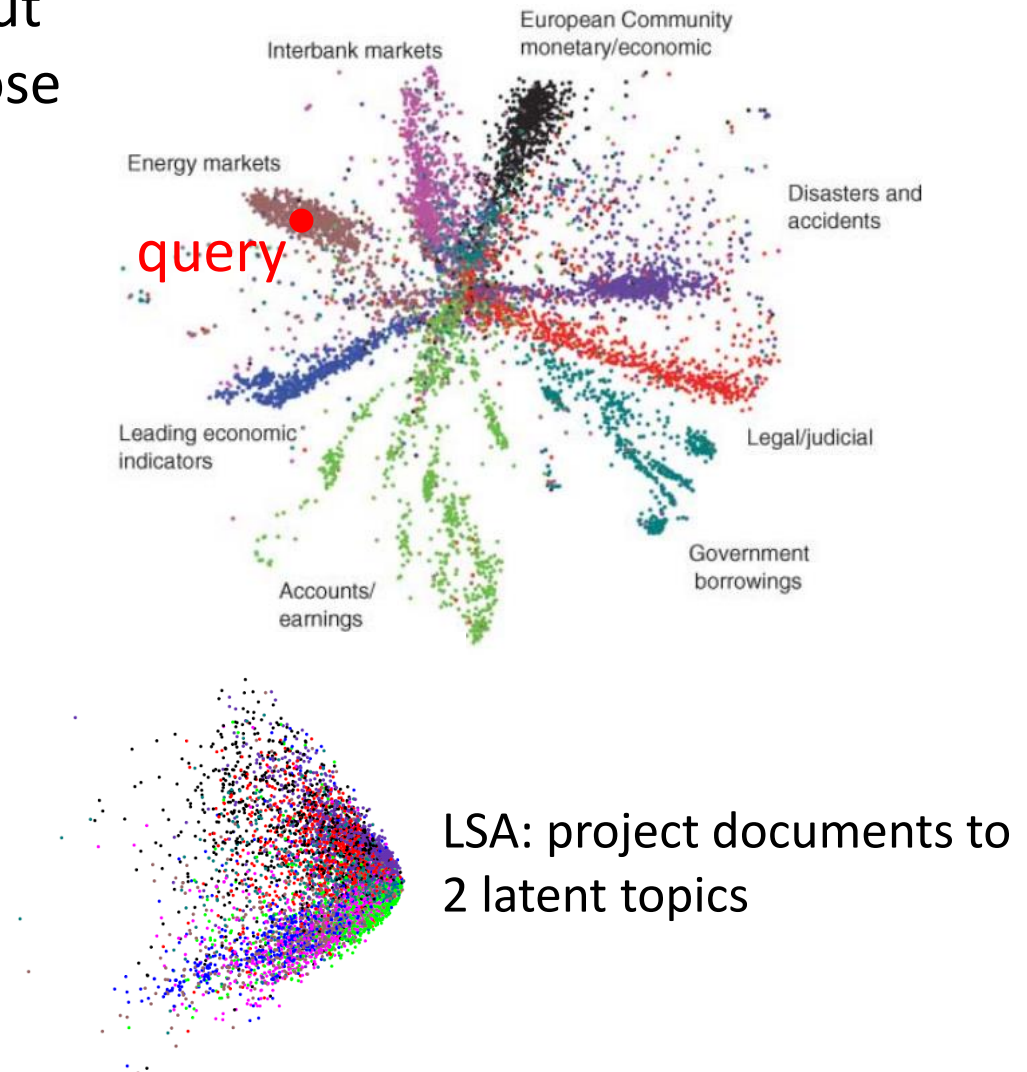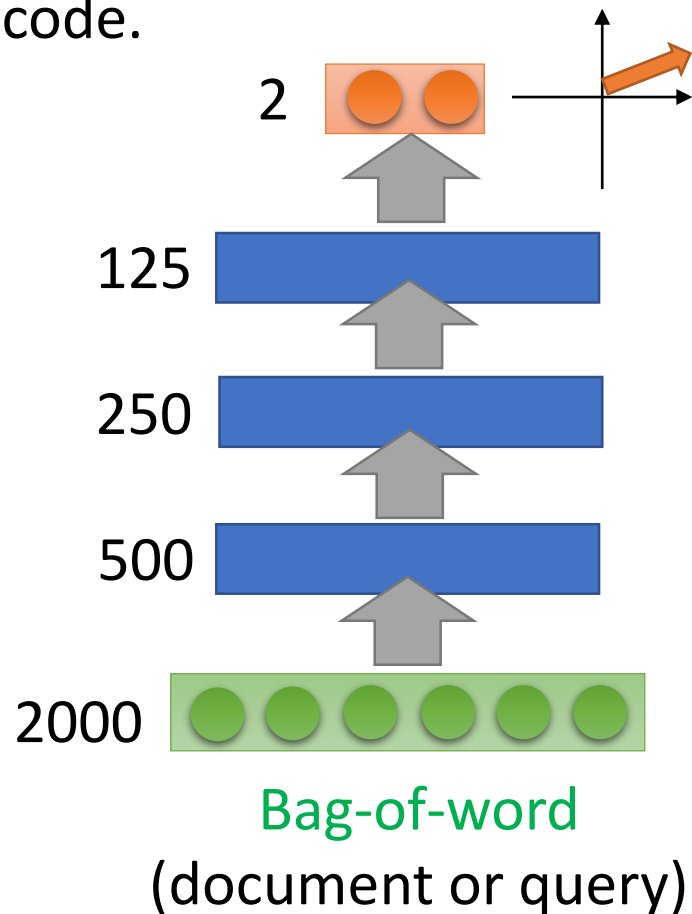| this | 1 |
is	1
a	0
an	1
apple	1
pen	0

Semantics are not considered.

Auto-encoder – Text Retrieval

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

2

125

250

500

2000

Bag-of-word
(document or query)

Interbank markets

European Community monetary/economic

Energy markets

query

Disasters and accidents

Leading economic indicators

Legal/judicial

Accounts/earnings

Government borrowings

LSA: project documents to 2 latent topics

Auto-encoder – Similar Image Search

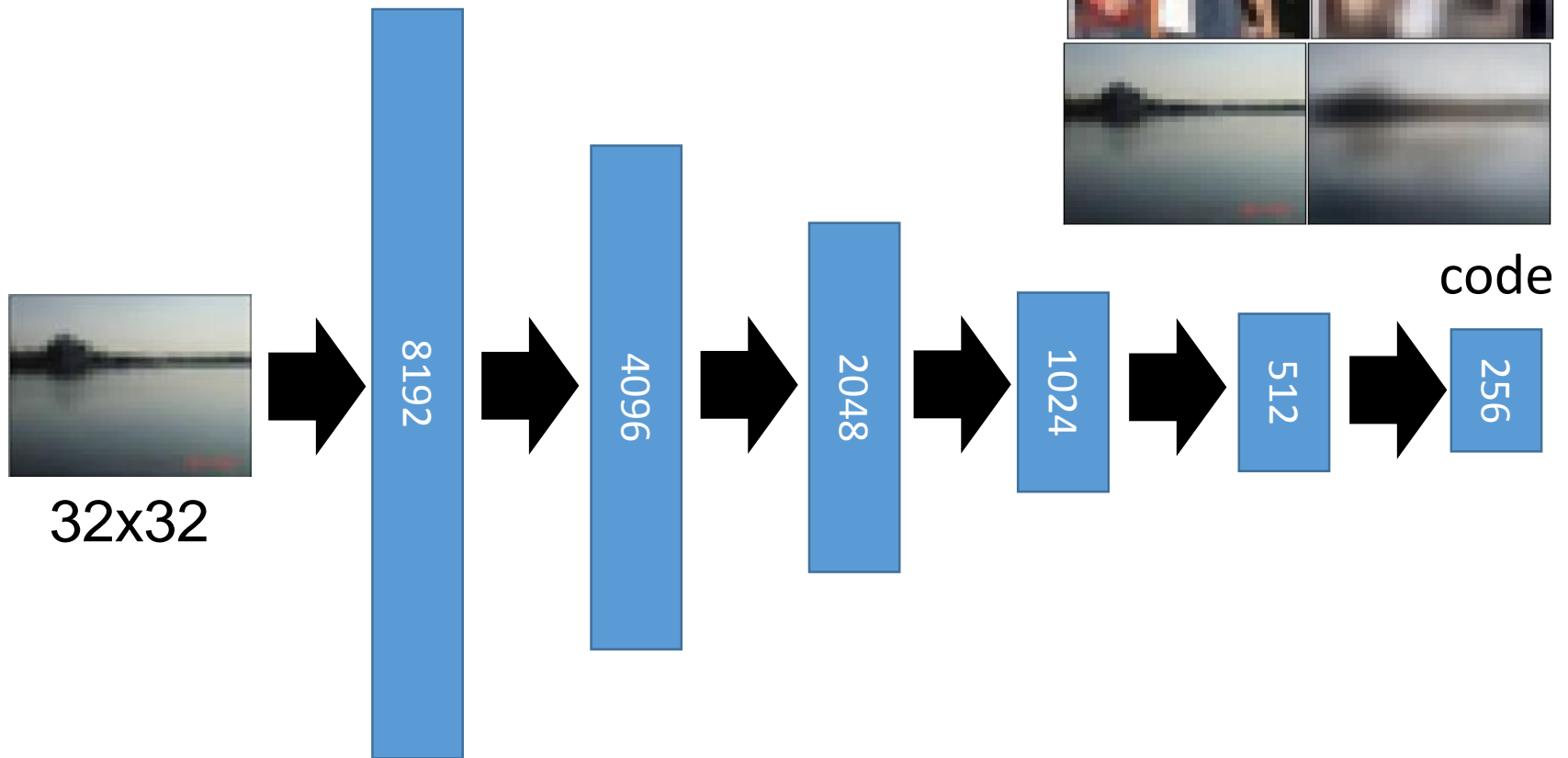Retrieved using Euclidean distance in pixel intensity space

(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.
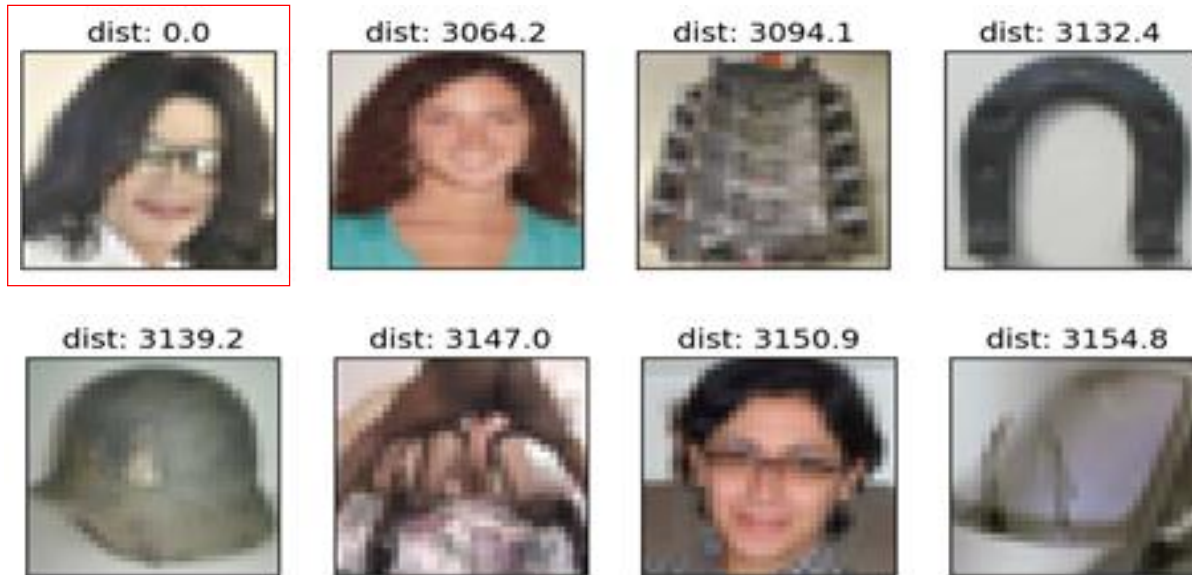
Auto-encoder – Similar Image Search

unsupervised   data

32x32

8192 → 4096 → 2048 → 1024 → 512 → 256

code

(crawl millions of images from the Internet)

Retrieved using Euclidean distance in pixel intensity space

dist: 0.0   dist: 3064.2   dist: 3094.1   dist: 3132.4

dist: 3139.2   dist: 3147.0   dist: 3150.9   dist: 3154.8

retrieved using 256 codes

class

dist: 0   dist: 61   dist: 64   dist: 65

dist: 66   dist: 67   dist: 67   dist: 67

ge Search >

Auto-encoder for CNN

As close as possible

Deconvolution	🐱
↑	↓
Unpooling	Convolution
↑	↓
Deconvolution	Pooling
↑	↓
Unpooling	Convolution
↑	↓
Deconvolution ← code ←	Pooling

CNN -Unpooling

14 x 14   28 x 28

Layer Above Reconstruction

Unpooling

Max Loc "Switc

Unpooled Maps

Max Locations "Switches"

Pooled Maps

Pooling

Rectified Feature Maps

Alternative: simply repeat the values

Source of image : https://leonardoaraujosantos.gitbooks.io/artificial-inteligence/content/image_segmentation.html
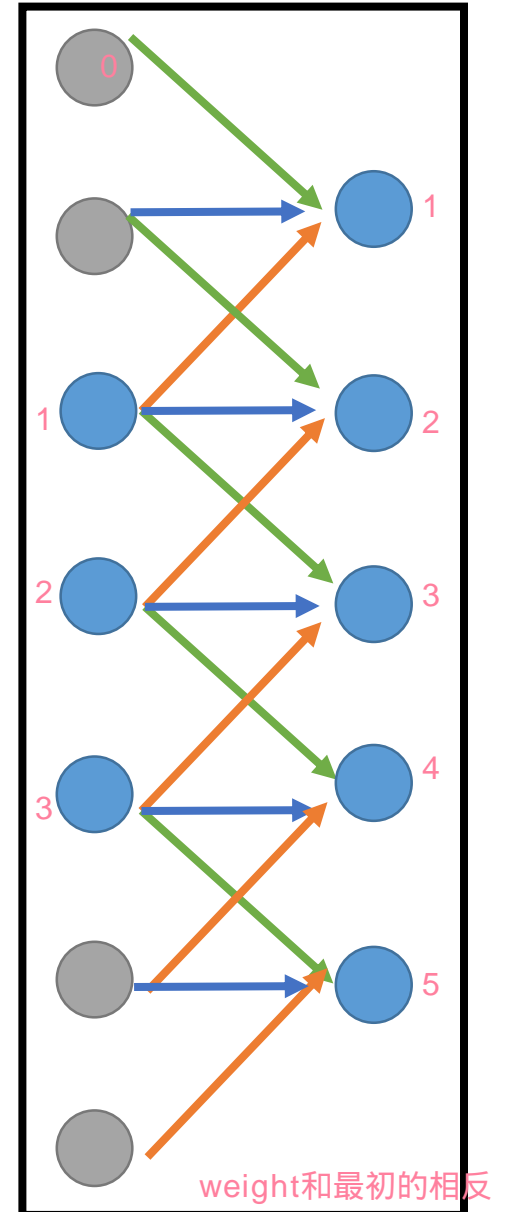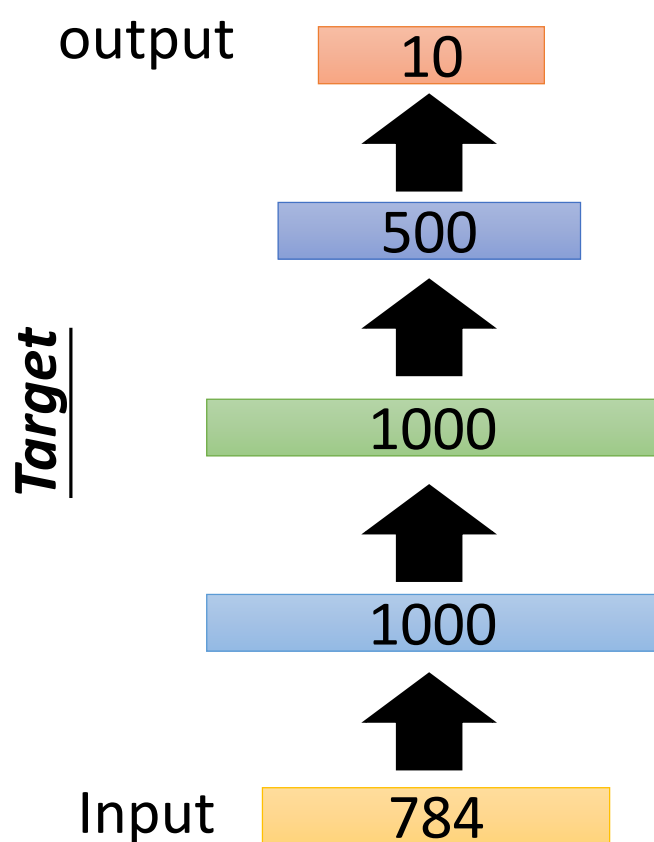
CNN
- Deconvolution

convolution

deconvolution

weight

Auto-encoder – Pre-training DNN

- Greedy Layer-wise Pre-training *again*

output

10

500

Target

1000

1000

Input   784

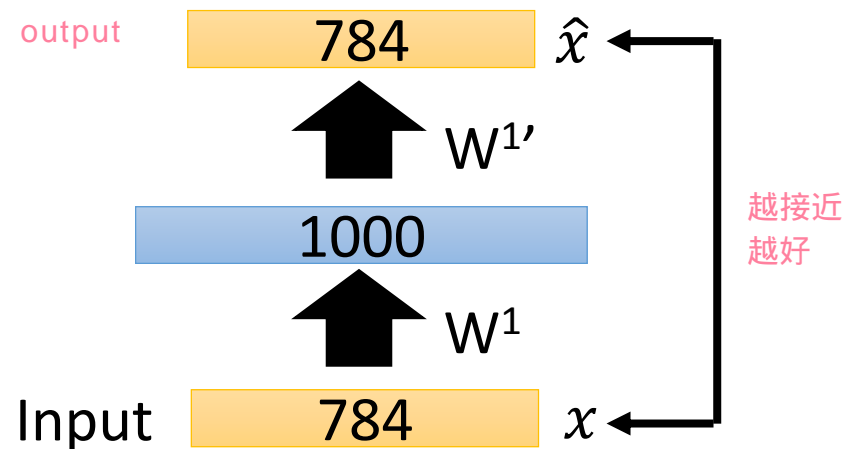code   dimension

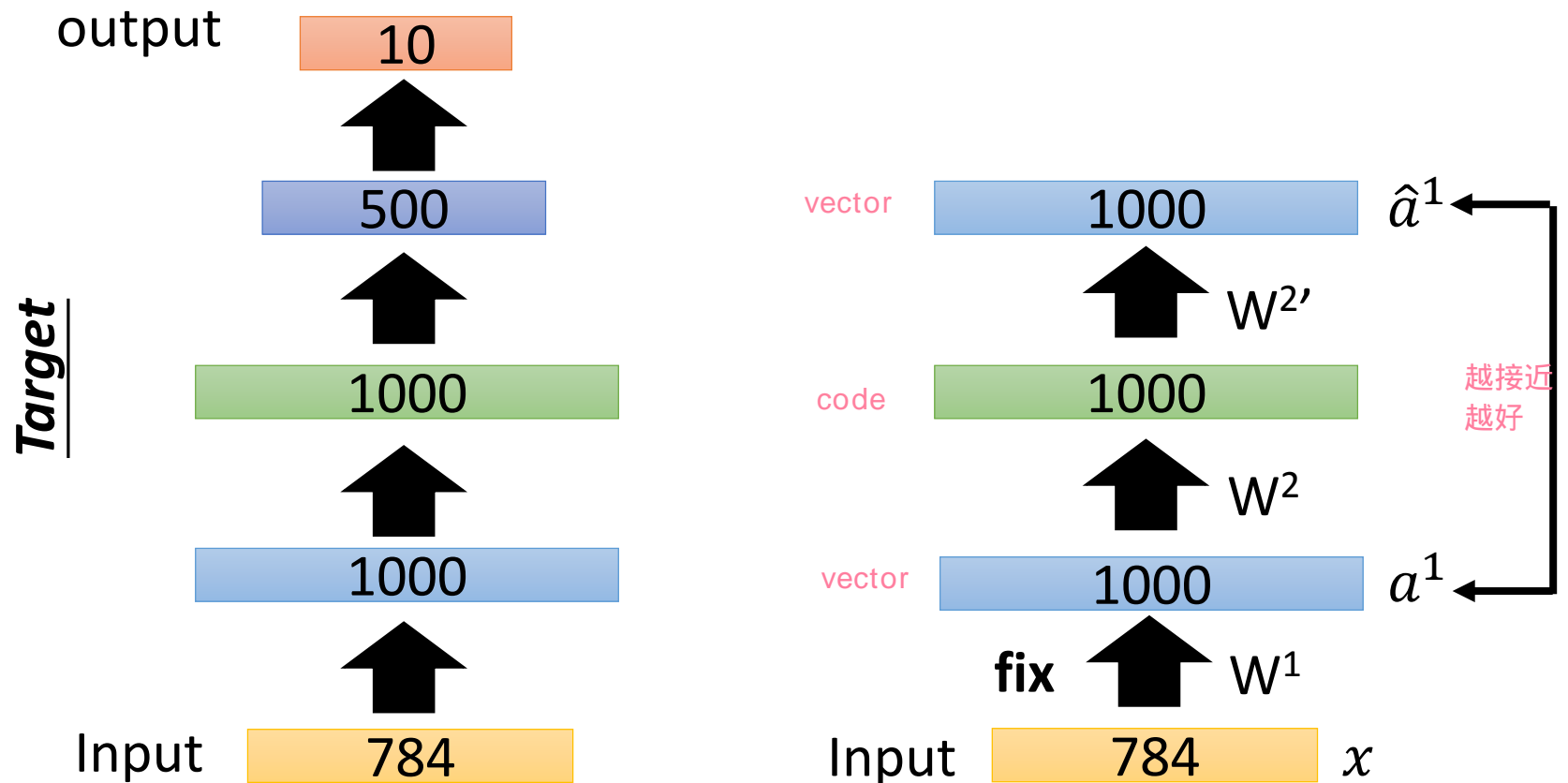output   784 $\hat{x}$
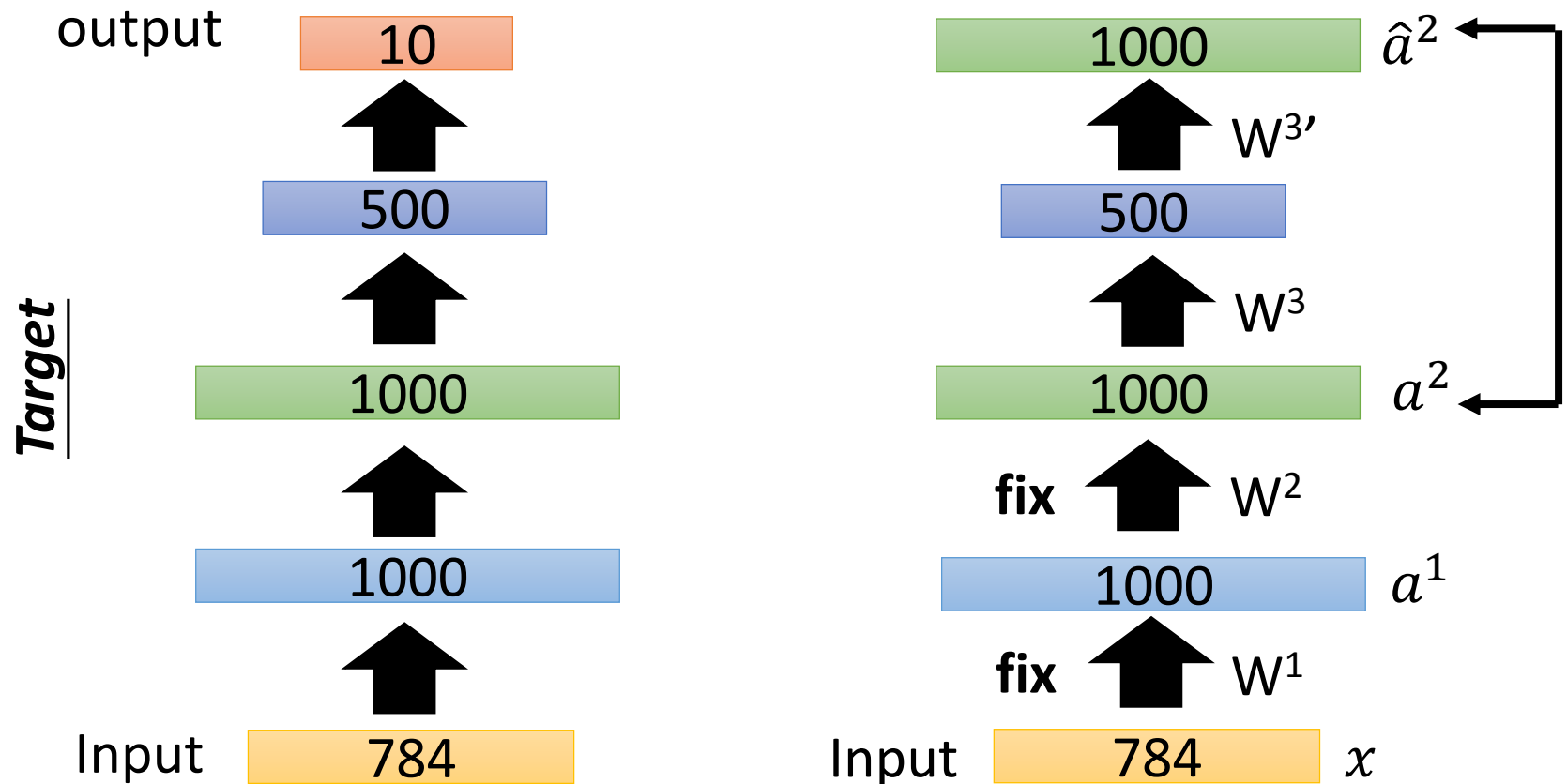
$W^{1\prime}$

1000

W^1

Input   784 x

Auto-encoder – Pre-training DNN

- Greedy Layer-wise Pre-training *again*

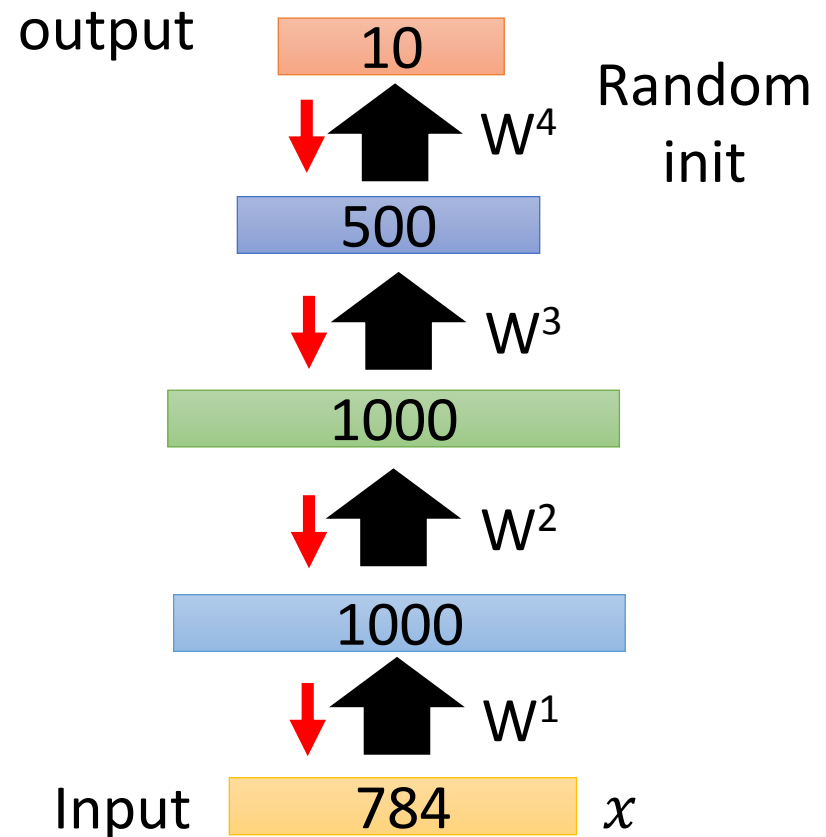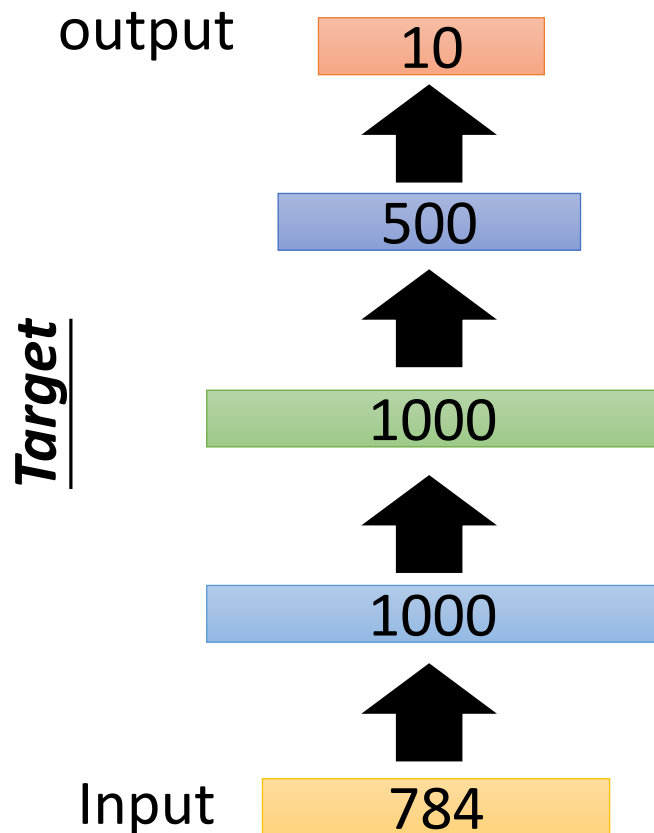# Auto-encoder – Pre-training DNN

- Greedy Layer-wise Pre-training *again*

Auto-encoder – Pre-training DNN

unlabel data

- Greedy Layer-wise Pre-training *again*

Find-tune by backpropagation

output | 10

Target

output | 10 | Random init | W^4 | W^3 | W^2 | W^1

500

1000

1000

Input | 784

500

1000

1000

Input | 784 | x

Learning More
- Restricted Boltzmann Machine

- Neural networks [5.1] : Restricted Boltzmann machine – definition
  - https://www.youtube.com/watch?v=p4Vh_zMw-HQ&index=36&list=PL6Xpj9I5qXYEcOhn7TqghAJ6NAPrNmUBH

- Neural networks [5.2] : Restricted Boltzmann machine – inference
  - https://www.youtube.com/watch?v=lekCh_i32iE&list=PL6Xpj9I5qXYEcOhn7TqghAJ6NAPrNmUBH&index=37

- Neural networks [5.3] : Restricted Boltzmann machine - free energy
  - https://www.youtube.com/watch?v=e0Ts_7Y6hZU&list=PL6Xpj9I5qXYEcOhn7TqghAJ6NAPrNmUBH&index=38
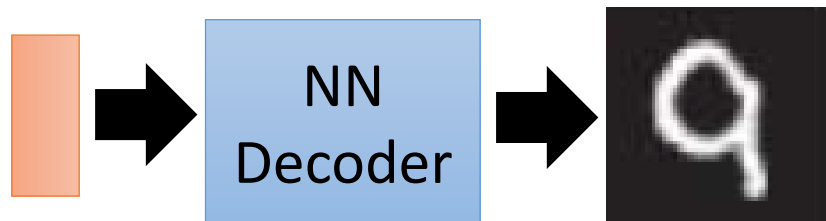
Learning More
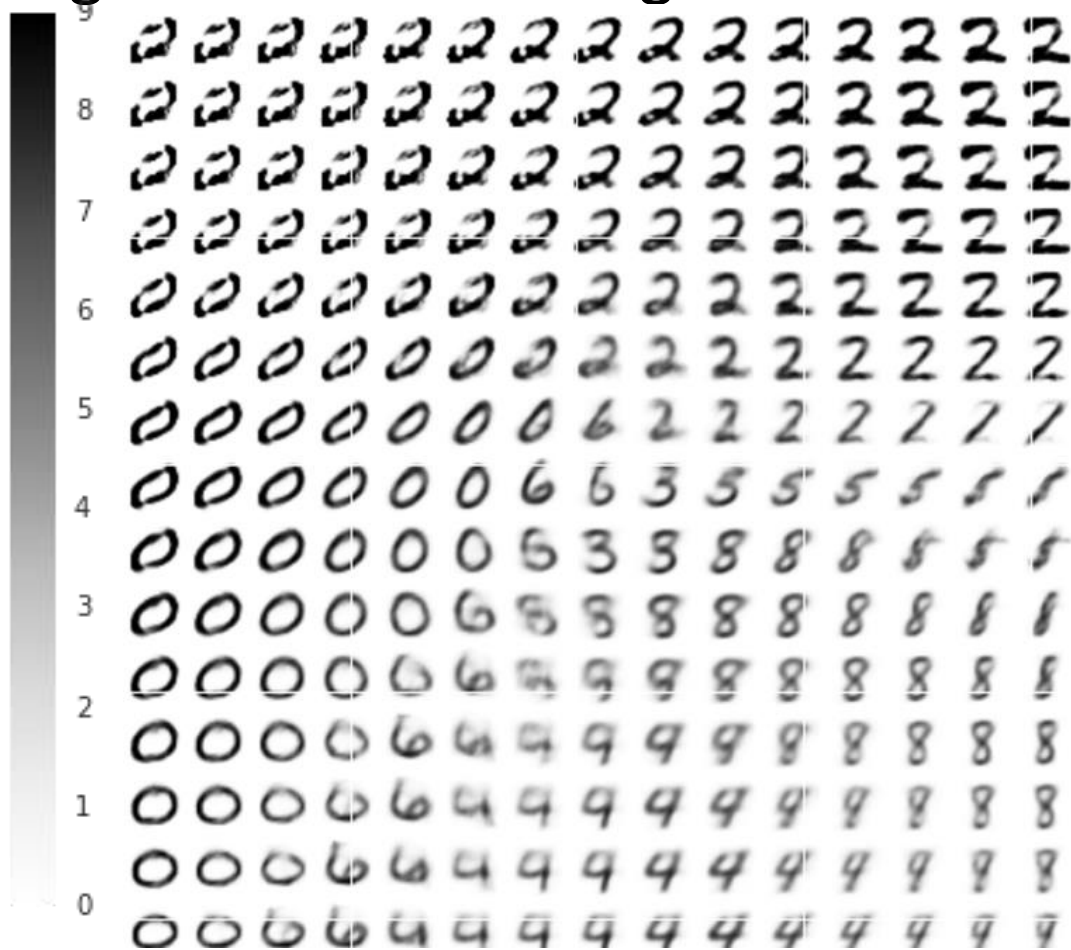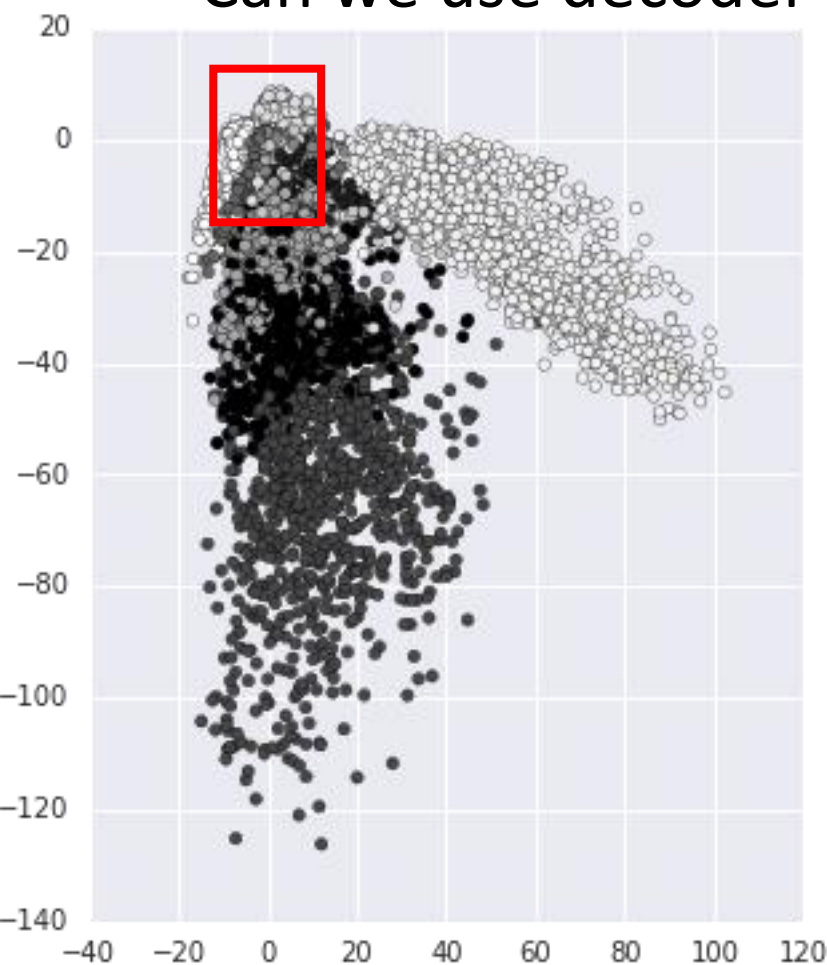# - Deep Belief Network DL

- Neural networks [7.7] : Deep learning - deep belief network
  - https://www.youtube.com/watch?v=vkb6AWYXZ5I&list=PL6Xpj9I5qXYEcOhn7TqghAJ6NAPrNmUBH&index=57

- Neural networks [7.8] : Deep learning - variational bound
  - https://www.youtube.com/watch?v=pStDscJh2Wo&list=PL6Xpj9I5qXYEcOhn7TqghAJ6NAPrNmUBH&index=58

- Neural networks [7.9] : Deep learning - DBN pre-training
  - https://www.youtube.com/watch?v=35MUlYCColk&list=PL6Xpj9I5qXYEcOhn7TqghAJ6NAPrNmUBH&index=59

Next …..
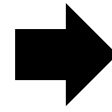

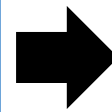NN Decoder

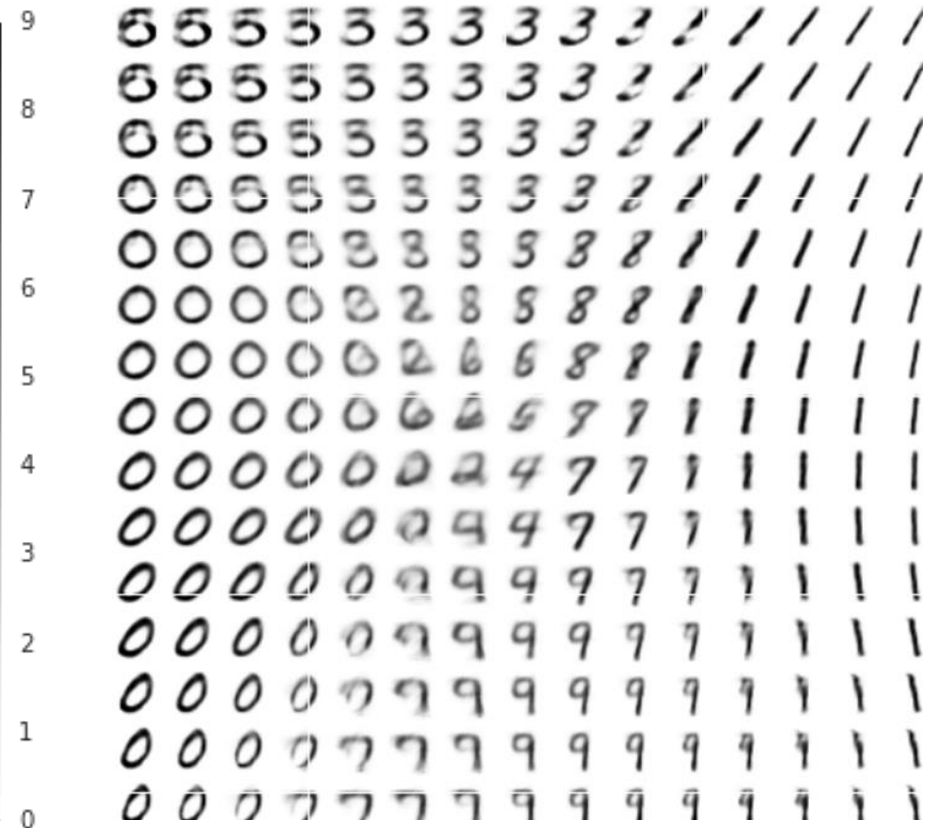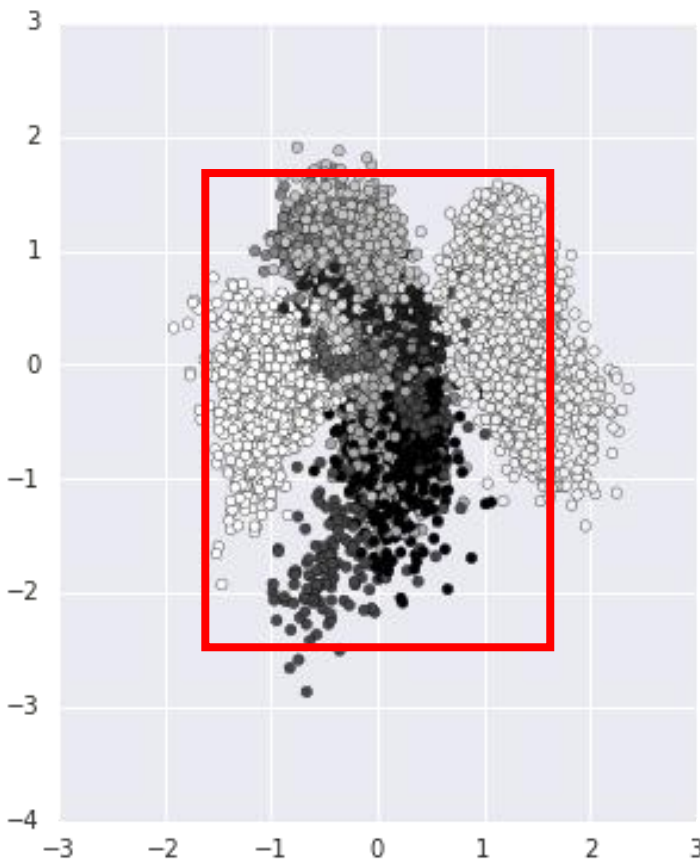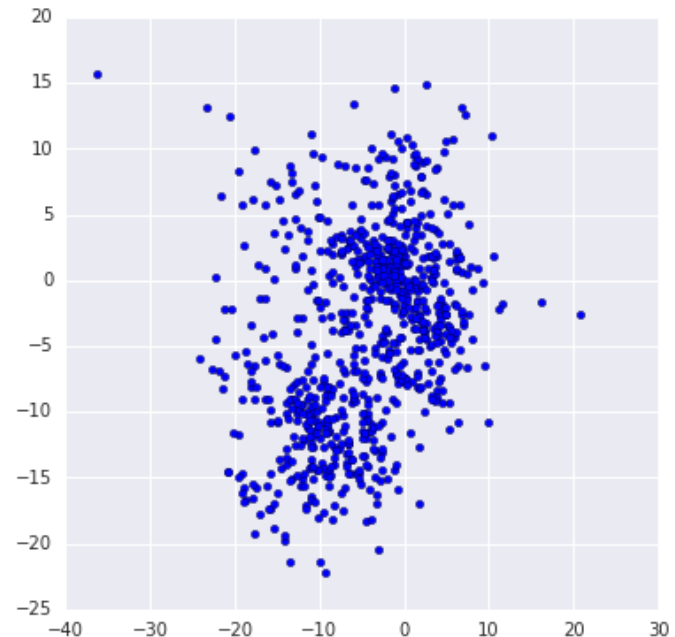- Can we use decoder to generate something?

Next …..

code → NN Decoder →

- Can we use decoder to generate something?

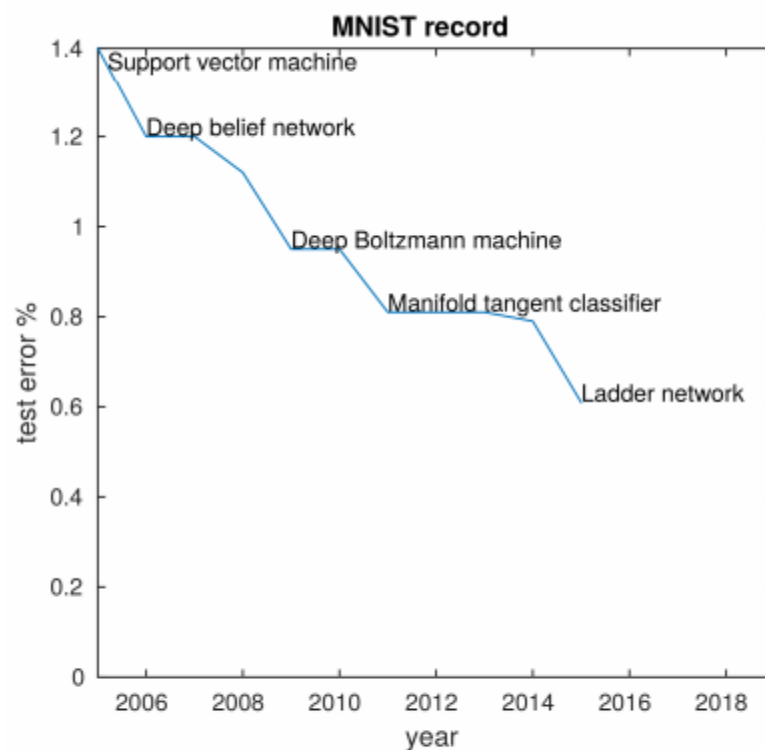# Appendix

Pokémon

- http://140.112.21.35:2880/~tlkagk/pokemon/pca.html
- http://140.112.21.35:2880/~tlkagk/pokemon/auto.html
- The code is modified from
  - http://jkunst.com/r/pokemon-visualize-em-all/

Add: Ladder Network

- http://rinuboney.github.io/2016/01/19/ladder-network.html
- https://mycourses.aalto.fi/pluginfile.php/146701/mod_resource/content/1/08%20semisup%20ladder.pdf
- https://arxiv.org/abs/1507.02672

Yearly progress in permutation-invariant MNIST.
A. Rasmus, H. Valpola, M. Honkala, M. Berglund, and T. Raiko.

Semi-Supervised Learning with Ladder Network. To appear in NIPS 2015.