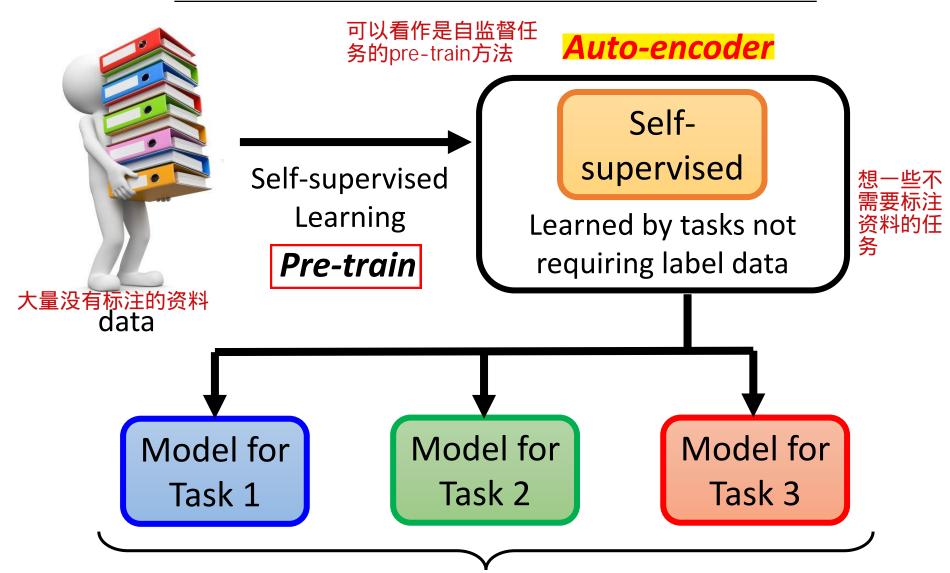
P53- Auto-encoder (上)

2021年5月1日 自编码器(1)-基本概念

# AUTO-ENCODER

Hung-yi Lee 李宏毅

### Self-supervised Learning Framework



**Downstream Tasks** 

#### Outline

Basic Idea of Auto-encoder

Feature Disentanglement

Discrete Latent Representation

More Applications

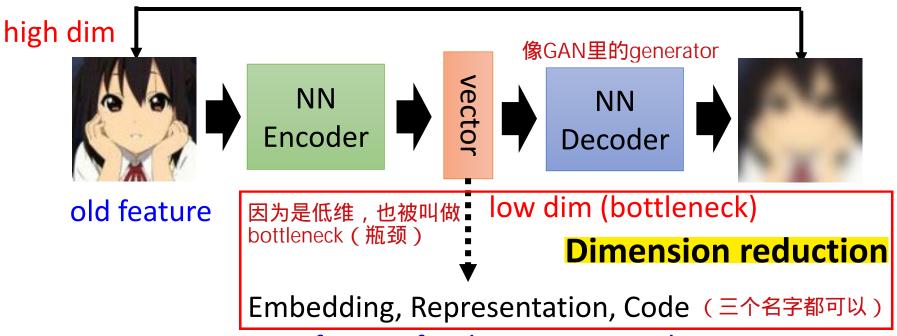
### Auto-encoder

Unlabeled Images



Sounds familiar? We have seen the same idea in Cycle GAN. ©

训练目标 As close as possible (reconstruction)



New feature for downstream tasks

### More Dimension Reduction

(not based on deep learning)



https://youtu.be/iwh5o\_M4BNU

PCA <sub>P59</sub>



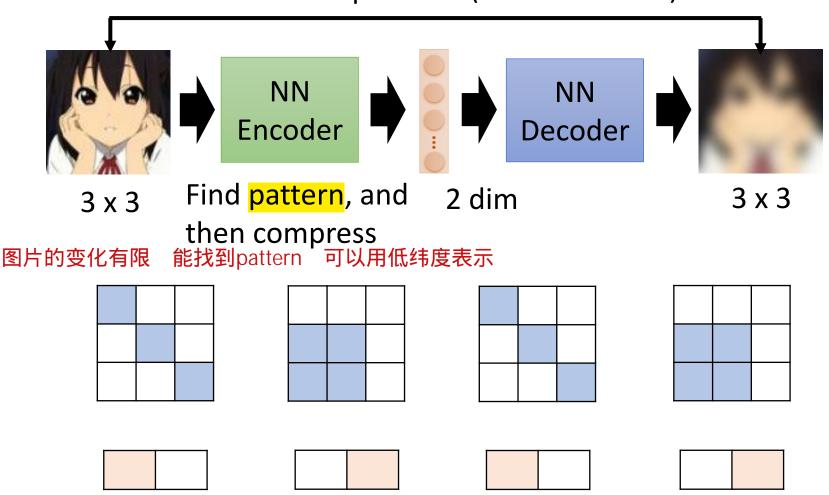
https://youtu.be/GBUEjkpoxXc

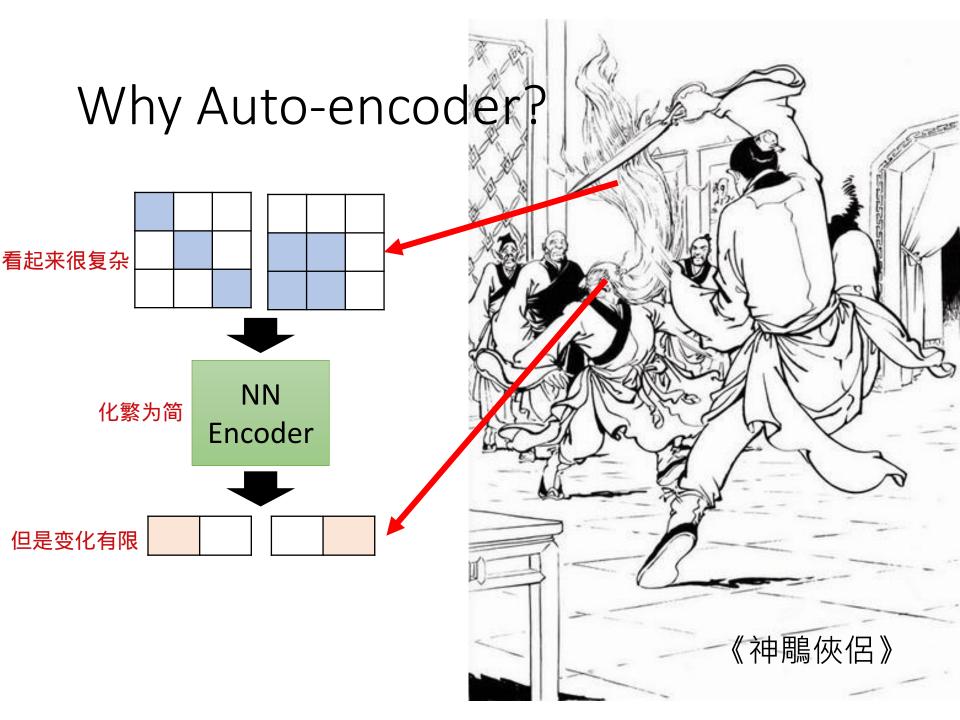
t-SNE

Why Auto-encoder? 《神鵰俠侶》

## Why Auto-encoder?

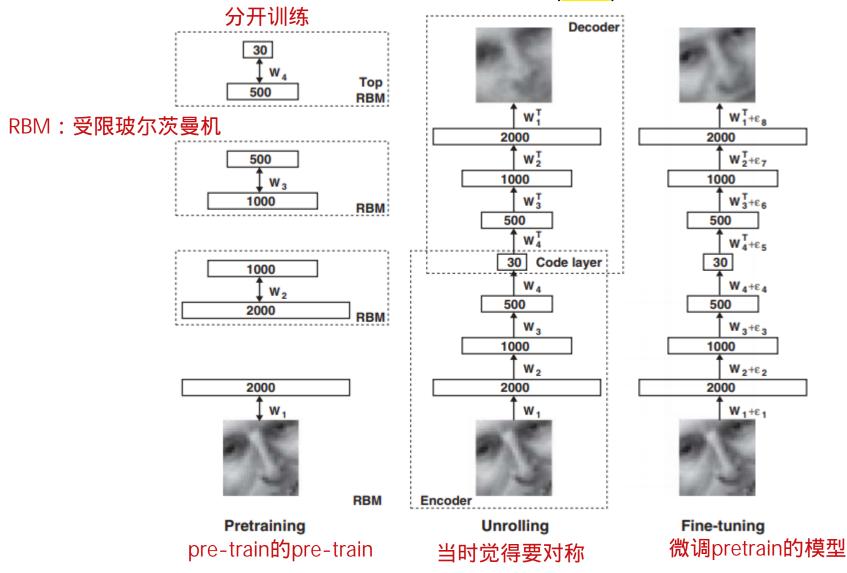
As close as possible (reconstruction)



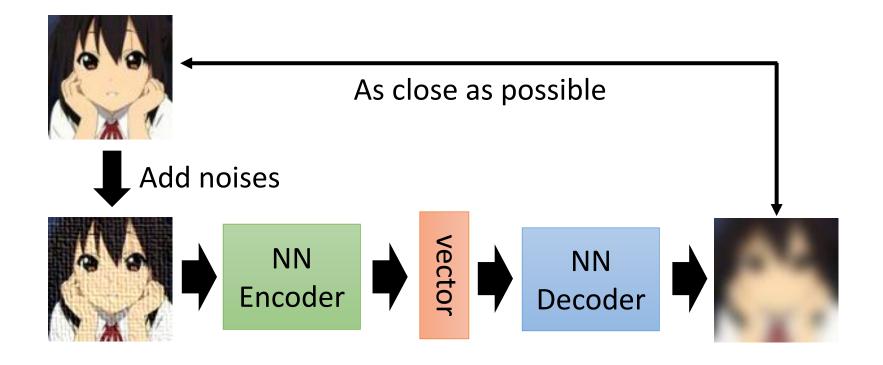


#### Auto-encoder is not a new idea

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



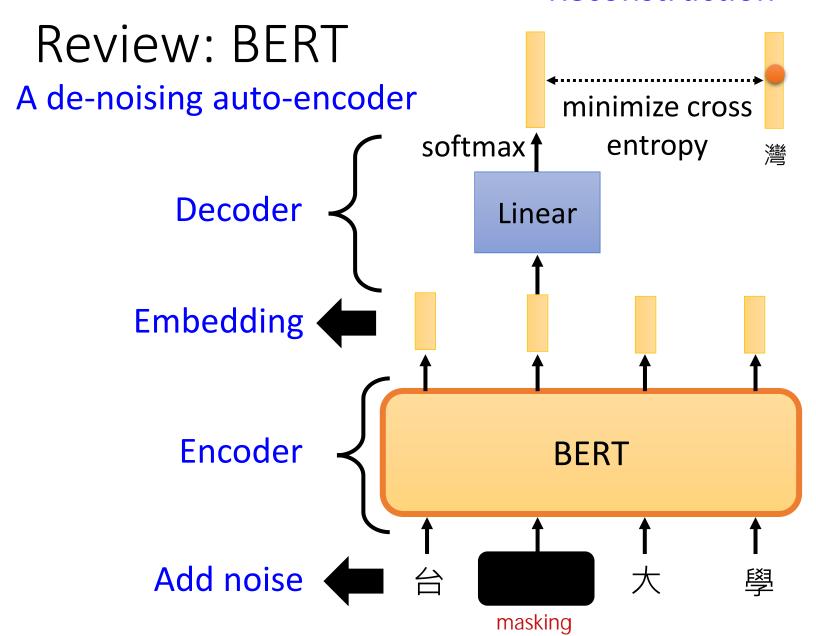
## De-noising Auto-encoder



The idea sounds familiar? © Bert

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

#### Reconstruction



### Outline

2020年5月1日 自编码器(2)

Basic Idea of Auto-encoder

Feature Disentanglement

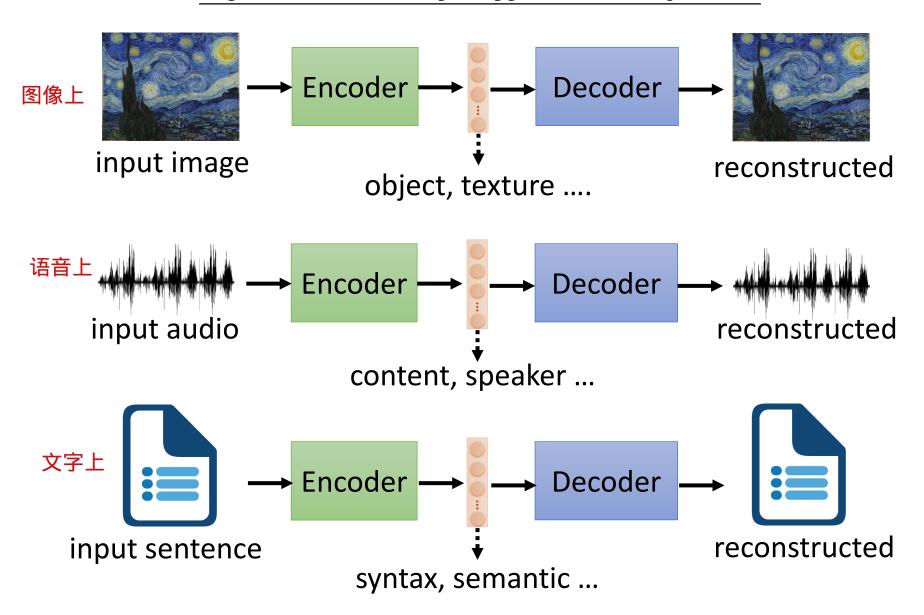
Discrete Latent Representation

More Applications

#### 什么是Disentanglement

#### Representation includes

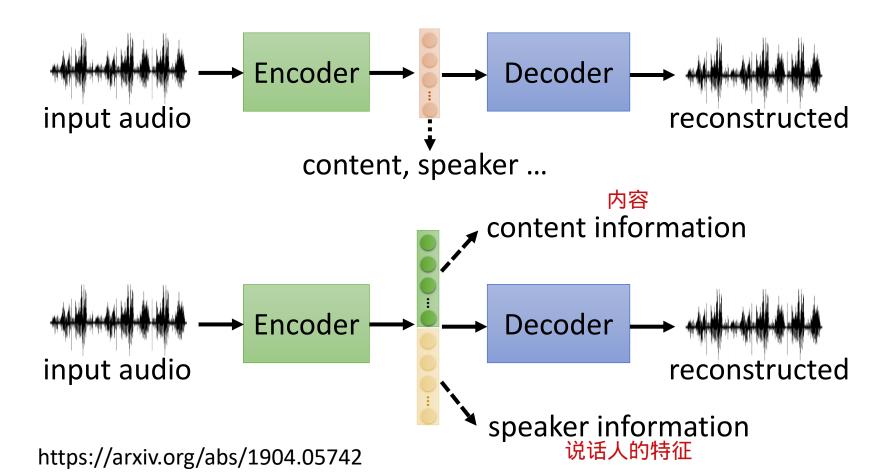
#### information of different aspects



### Feature Disentangle

https://arxiv.org/abs/1804.02812

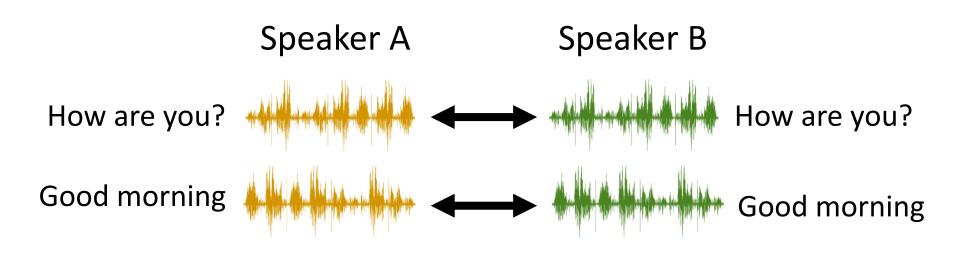
https://arxiv.org/abs/1905.05879

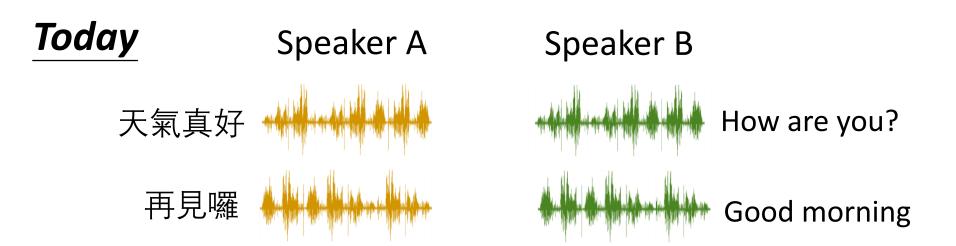




Application: Voice Conversion

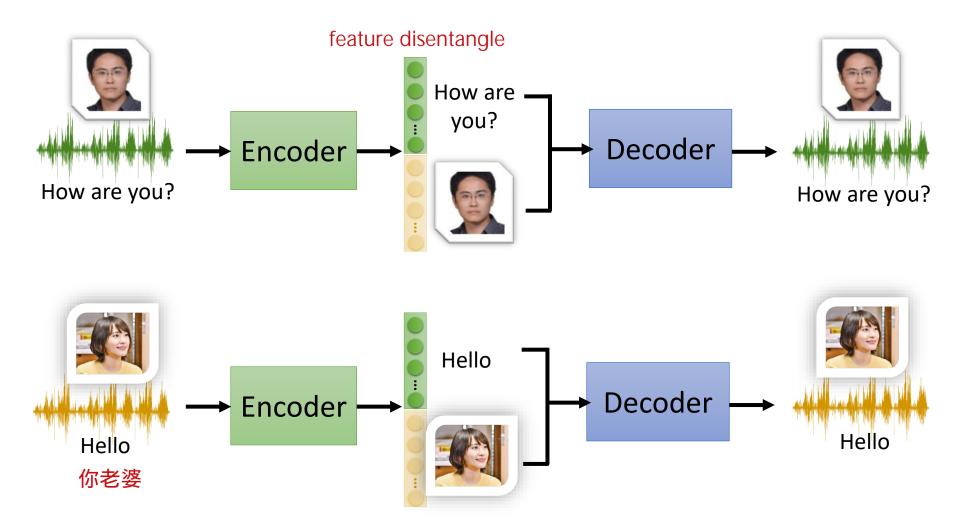
### In the past



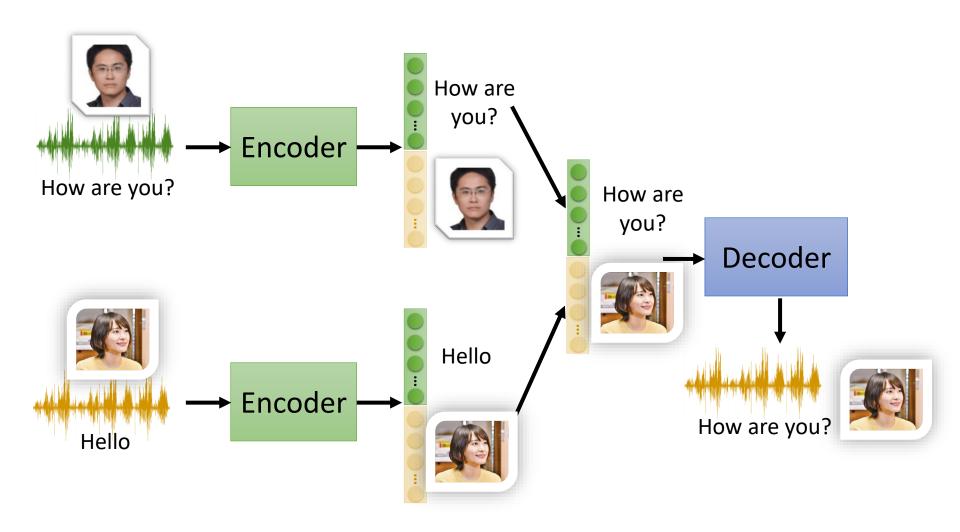


Speakers A and B are talking about completely different things.

## Application: Voice Conversion

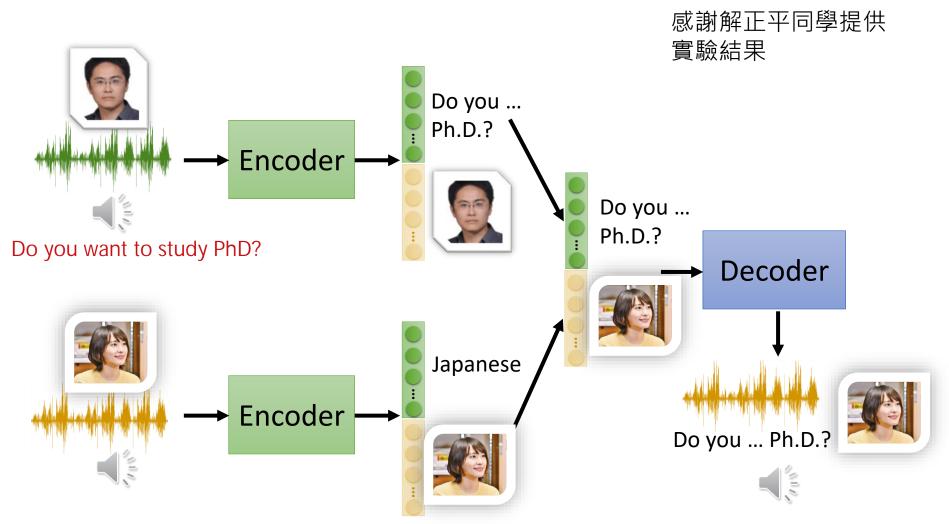


## Application: Voice Conversion



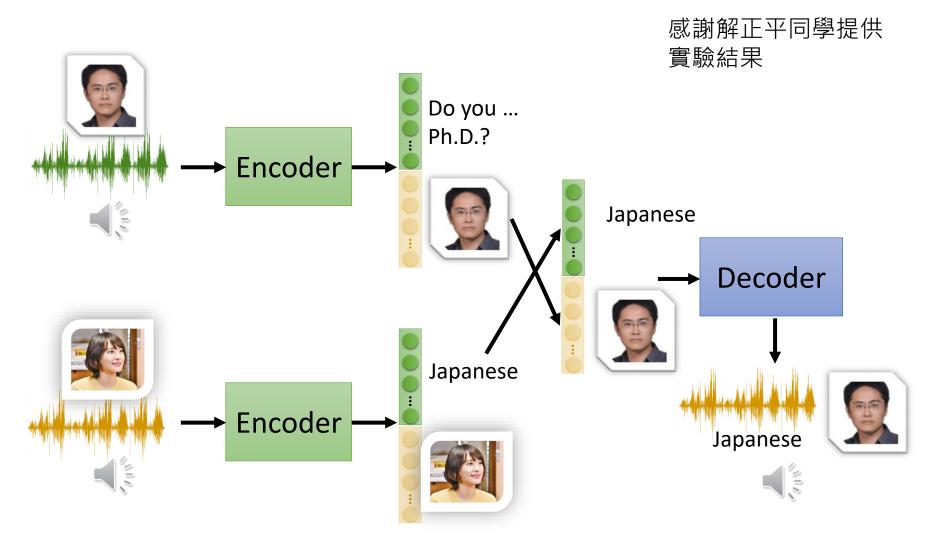
#### 实际操作

## Application: Voice Conversion



李宏毅:你老婆在叫你读博班!

## Application: Voice Conversion



### Outline

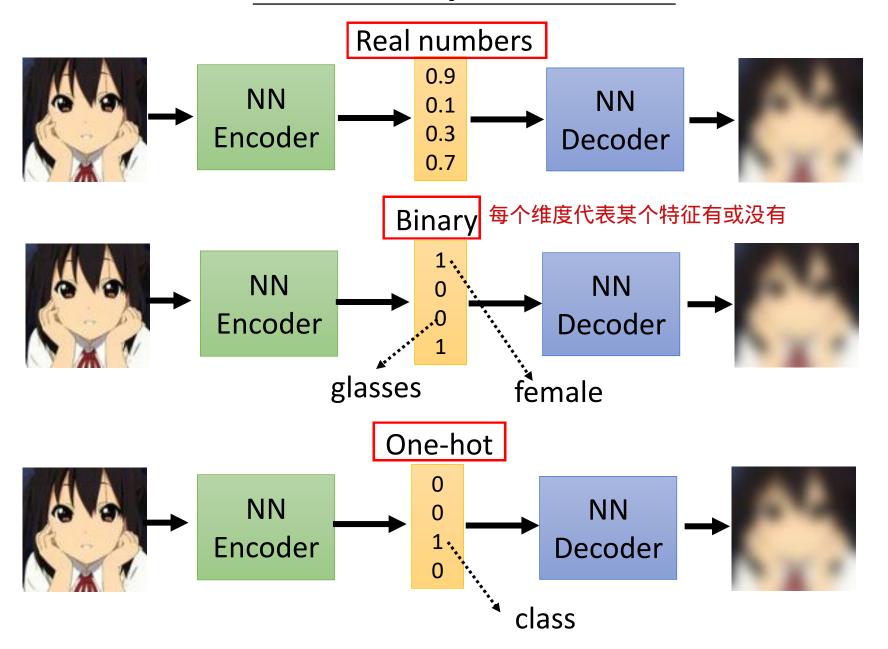
Basic Idea of Auto-encoder

Feature Disentanglement

Discrete Latent Representation

More Applications

#### Discrete Representation

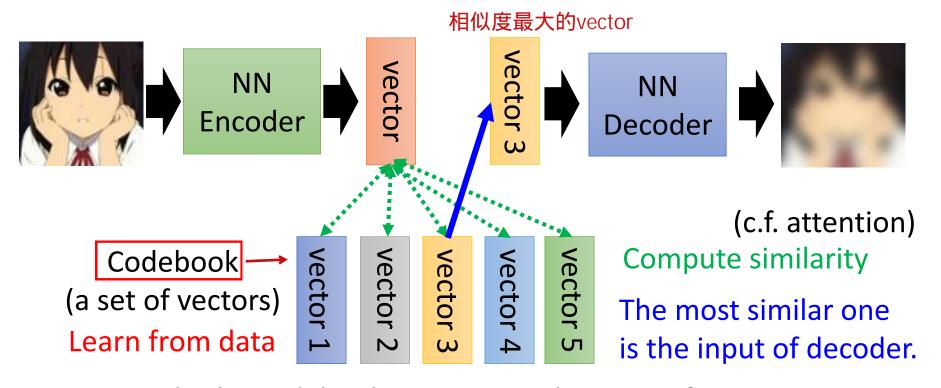


### Discrete Representation

最知名的Discrete Representation

https://arxiv.org/abs/1711.00937

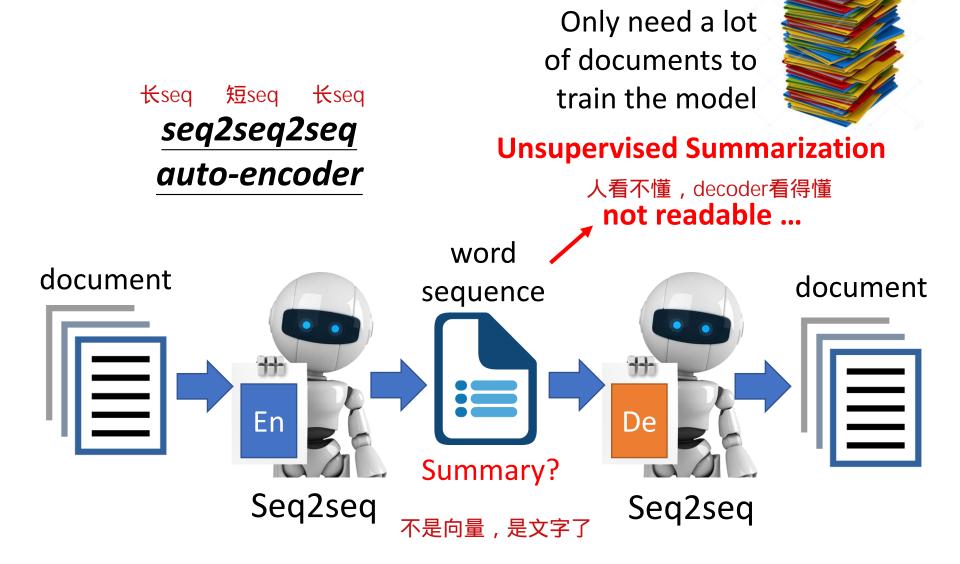
Vector Quantized Variational Auto-encoder (VQVAE)



For speech, the codebook represents phonetic information

https://arxiv.org/pdf/1901.08810.pdf

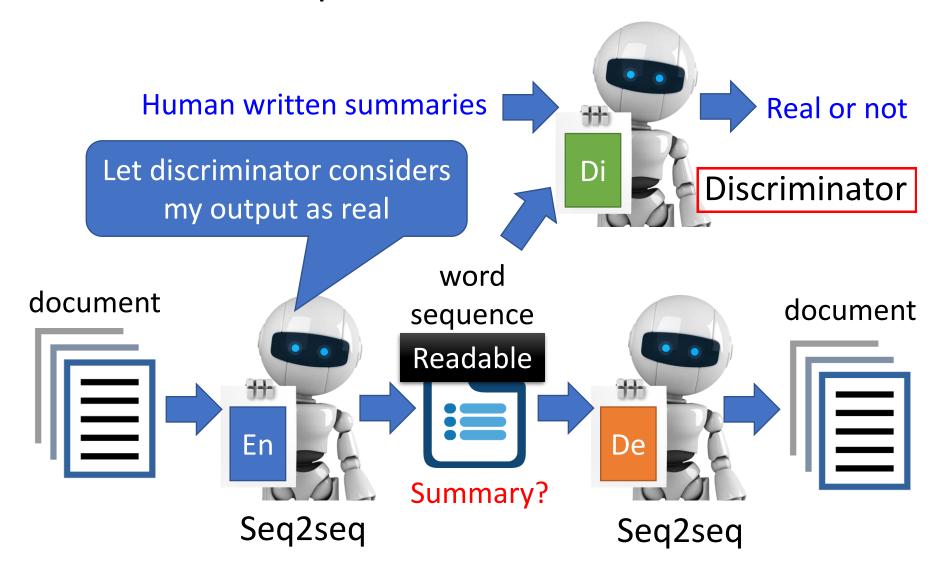
## Text as Representation



#### This is cycle GAN ©

### Text as Representation

RL硬train



## Text as Representation

• **Document**: 澳大利亞今天與13個國家簽署了反興奮劑雙邊協議, 旨在加強體育競賽之外的藥品檢查並共享研究成果 ......

#### • Summary:

- Human: 澳大利亞與13國簽署反興奮劑協議
- Unsupervised: 澳大利亞加強體育競賽之外的藥品檢查
- **Document**:中華民國奧林匹克委員會今天接到一九九二年 冬季奧運會邀請函,由於主席張豐緒目前正在中南美洲進 行友好訪問,因此尚未決定是否派隊赴賽 ......

#### Summary:

- Human:一九九二年冬季奧運會函邀我參加
- Unsupervised: 奧委會接獲冬季奧運會邀請函

## Text as Representation

• **Document**:據此間媒體27日報道,印度尼西亞蘇門答臘島的兩個省近日來連降暴雨,洪水泛濫導致塌方,到26日為止至少已有60人喪生,100多人失蹤 ......

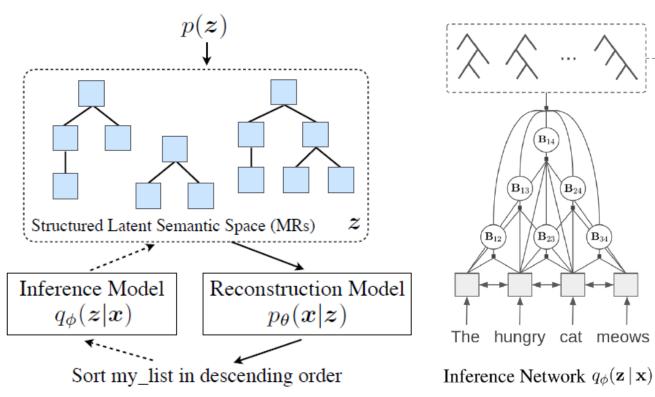
#### • Summary:

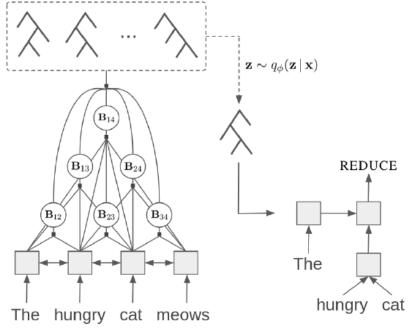
- Human:印尼水災造成60人死亡
- Unsupervised:印尼門洪水泛濫導致塌雨 错误例子
- **Document**:安徽省合肥市最近為領導幹部下基層做了新規 定:一律輕車簡從,不準搞迎來送往、不準搞層層陪同 .....

#### Summary:

- Human:合肥規定領導幹部下基層活動從簡
- Unsupervised: 合肥領導幹部下基層做搞迎來送往規定:
  - 一律簡

## Tree as Embedding





https://arxiv.org/abs/1806.07832

https://arxiv.org/abs/1904.03746

Generative Model  $p_{\theta}(\mathbf{x}, \mathbf{z})$ 

### Outline

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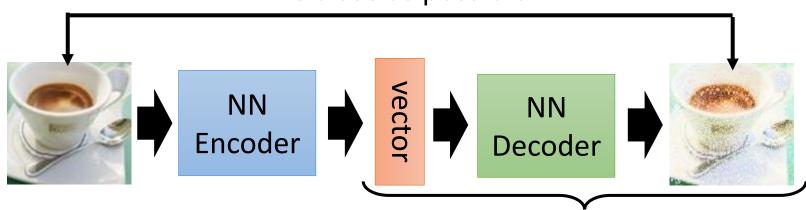
**More Applications** 

### Generator

generative模型

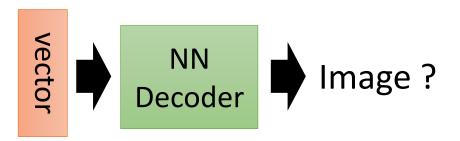
- 1. GAN
- 2. VAE
- 3.

#### As close as possible



Is it a generator?

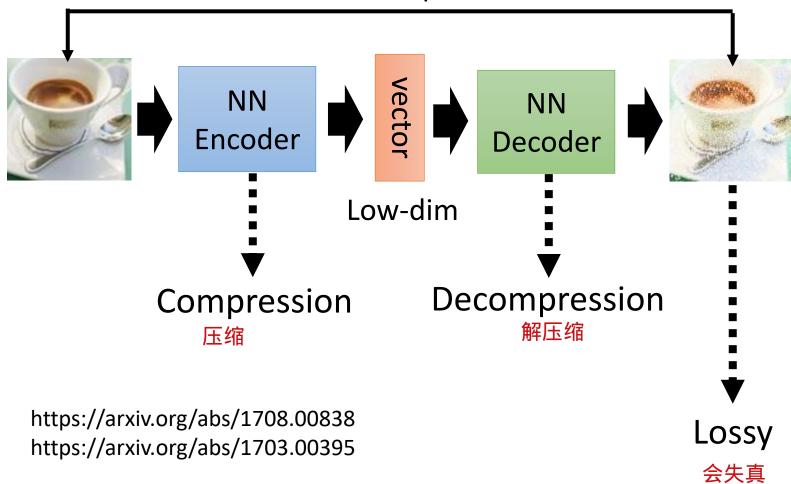
Randomly generate a vector from a distribution



With some modification, we have variational auto-encoder (VAE).

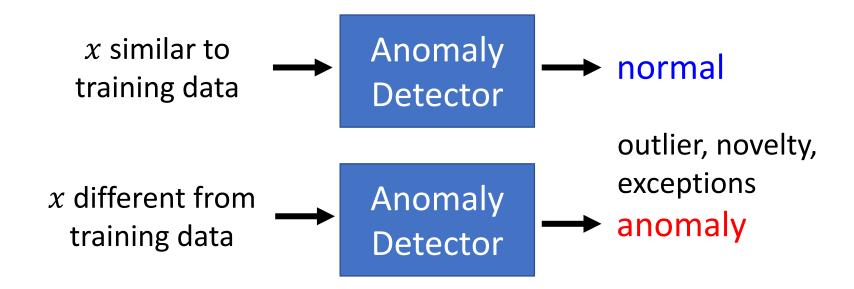
### Compression





### Anomaly Detection 异常检测

- Given a set of training data  $\{x^1, x^2, \dots, x^N\}$
- Detecting input x is similar to training data or not.



## **Anomaly Detection**



宝可梦

难点:在于收资料,通常容易收到正常资料。 假设有大量的正常资料,一点点的异常资料, 所以不是一般的分类问题。

## Anomaly Detection

Binary Classification?
We only have one class.
Training auto-encoder

#### Fraud Detection

- Training data: credit card transactions, x: fraud or not
- Ref: https://www.kaggle.com/ntnu-testimon/paysim1/home
- Ref: https://www.kaggle.com/mlg-ulb/creditcardfraud/home

#### Network Intrusion Detection

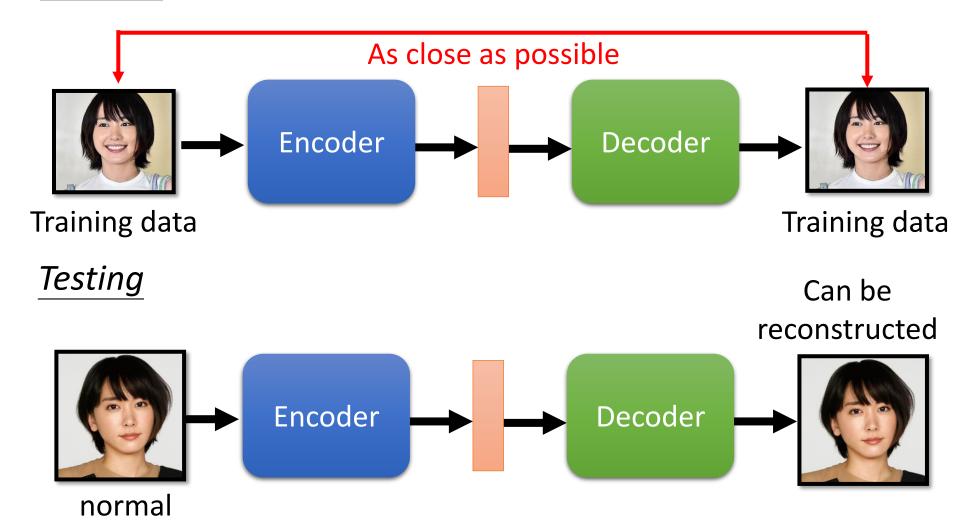
- Training data: connection, x: attack or not
- Ref: http://kdd.ics.uci.edu/databases/kddcup99/kddcup99.html

#### Cancer Detection

- Training data: normal cells, x: cancer or not?
- Ref: https://www.kaggle.com/uciml/breast-cancer-wisconsin-data/home

#### Approach: Auto-encoder

Training Using real human faces to learn an autoencoder



#### Approach: Auto-encoder

**Training** Using **real human faces** to learn an **autoencoder** As close as possible Encoder Decoder Training data Training data Testing cannot be Large reconstruction loss → anomaly reconstructed Encoder Decoder

凉宫春日,二次元人物

anomaly

## More about Anomaly Detection

- Part 1: https://youtu.be/gDp2LXGnVLQ
- Part 2: https://youtu.be/cYrNjLxkoXs
- Part 3: https://youtu.be/ueDlm2FkCnw
- Part 4: https://youtu.be/XwkHOUPbc0Q
- Part 5: https://youtu.be/Fh1xFBktRLQ
- Part 6: https://youtu.be/LmFWzmn2rFY
- Part 7: https://youtu.be/6W8FqUGYyDo

## Concluding Remarks

Basic Idea of Auto-encoder

Feature Disentanglement

Discrete Latent Representation

More Applications