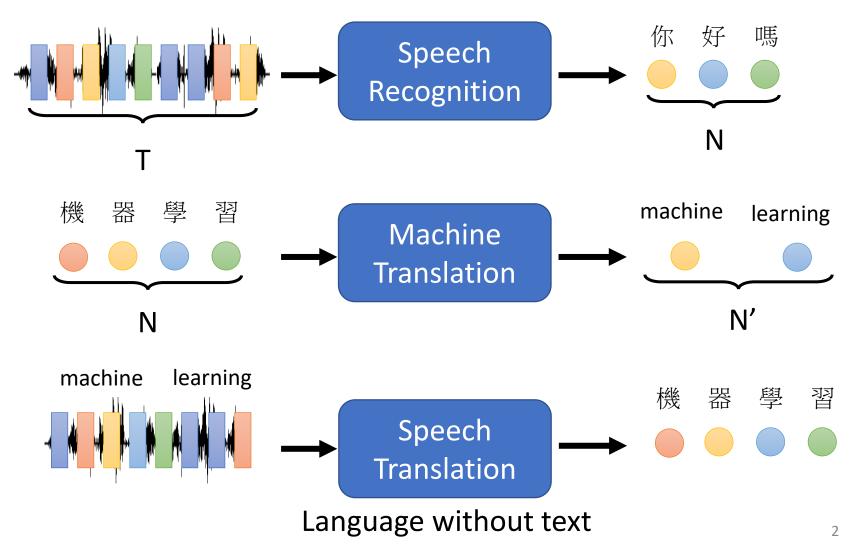


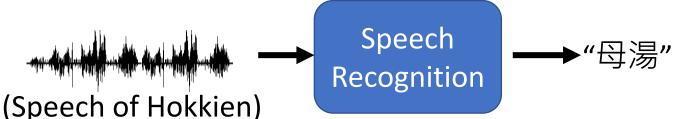
Sequence-to-sequence (Seq2seq)

Input a sequence, output a sequence

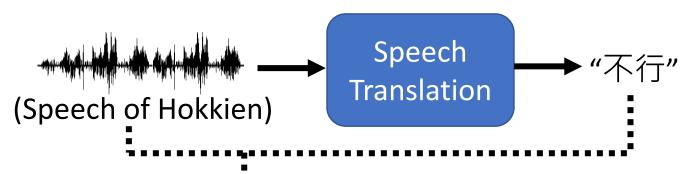
The output length is determined by model.



Hokkien (閩南語、台語)









Local soap operas (鄉土劇) on YouTube (Speech of Hokkien, Chinese subtitle)

Using 1500 hours of data for training



Hokkien (閩南語、台語)

Background music & noises?

Noisy transcriptions?

Phonemes of Hokkien?



"硬train—發" (Ying Train Yi Fa)

Hokkien (閩南語、台語)

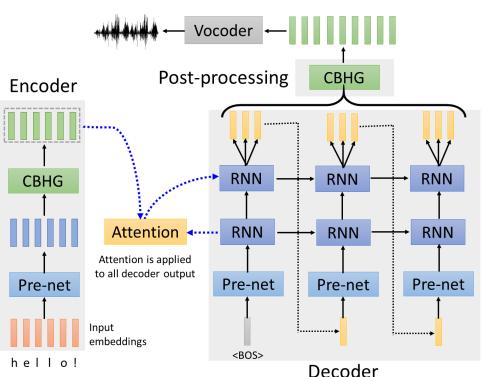
- 你的身體撐不住
- 沒事你為什麼要請假
- 要生了嗎 Answer:不會膩嗎
- 我有幫廠長拜託

Answer: 我拜託廠長了

To learn more: https://sites.google.com/speech.ntut.edu.tw/fsw/home/challenge-2020

Text-to-Speech (TTS) Synthesis

感謝張凱為同學提供實驗結果



Taiwanese Speech Synthesis

Source of data: 台灣媠聲2.0

歡迎來到台大語音處理實驗室



最近肺炎真嚴重,要記得戴口罩、 勤洗手,有病就要看醫生



Seq2seq for Chatbot

"Hello! How are you today?"

input seq2seq response

"Hi"

[PERSON 1:] Hi

Training

data:

[PERSON 2:] Hello! How are you today?

[PERSON 1:] I am good thank you, how are you.

[PERSON 2:] Great, thanks! My children and I were just about to watch Game of Thrones.

[PERSON 1:] Nice! How old are your children?

[PERSON 2:] I have four that range in age from 10 to 21. You?

[PERSON 1:] I do not have children at the moment.

[PERSON 2:] That just means you get to keep all the popcorn for yourself.

[PERSON 1:] And Cheetos at the moment!

[PERSON 2:] Good choice. Do you watch Game of Thrones?

[PERSON 1:] No, I do not have much time for TV.

[PERSON 2:] I usually spend my time painting: but, I love the show.

Most Natural Language Processing applications ...

Question Answering (QA)

Context Answer Question What is a major importance ...Southern California is a major major economic of Southern California in relation economic center for the state center to California and the US? of California and the US.... Der Großteil der What is the translation Most of the planet is from English to German? Erde ist Meerwasser ocean water. What is the Harry Potter star Daniel Harry Potter star summary? Radcliffe gains access to a Daniel Radcliffe gets reported £320 million fortune... £320M fortune... Hypothesis: Product and geography Premise: Conceptually cream are what make cream skimming skimming has two basic Entailment work. Entailment, neutral, dimensions - product and geography. or contradiction? A stirring, funny and finally transporting re-imagining of Is this sentence positive Beauty and the Beast and positive or negative? 1930s horror film. (sentiment analysis) decaNLP

QA can be done by seq2seq

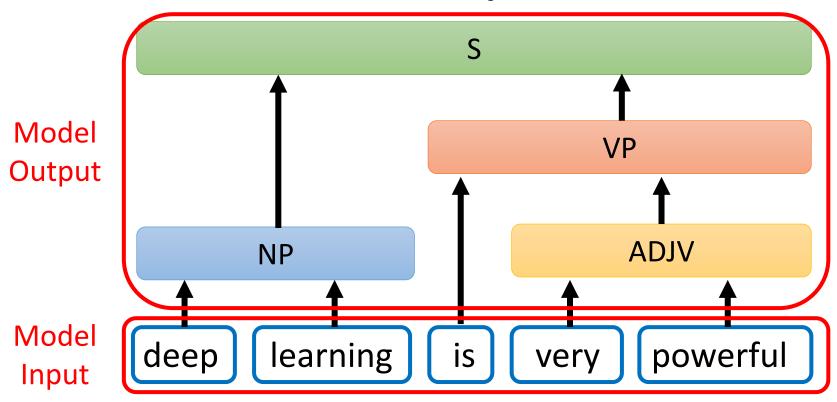
question, context
Seq2seq

answer

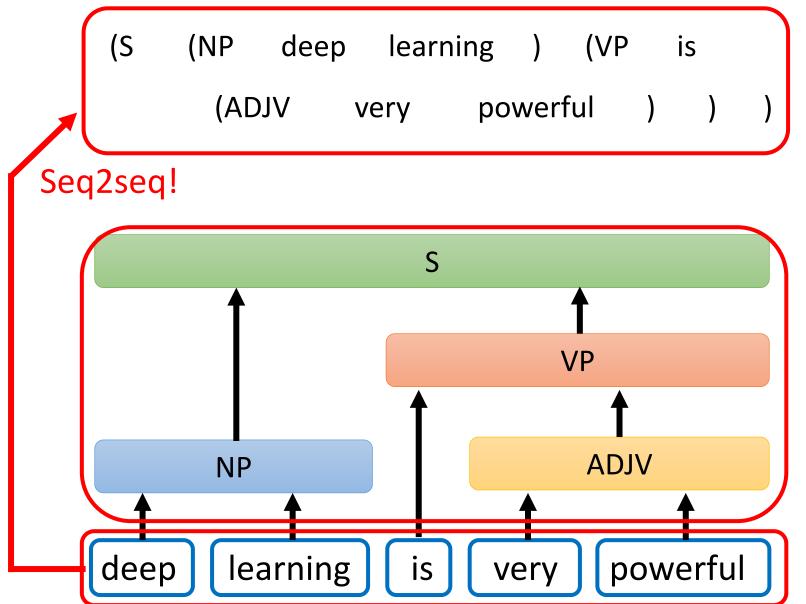
https://arxiv.org/abs/1806.08730 https://arxiv.org/abs/1909.03329

Seq2seq for Syntactic Parsing

Is it a sequence?



Seq2seq for Syntactic Parsing



Seq2seq for Syntactic Parsing

(S (NP deep learning) (VP is (ADJV very powerful))

Grammar as a Foreign Language

Oriol Vinyals*
Google
vinyals@google.com

Terry Koo

Google

terrykoo@google.com

Lukasz Kaiser*
Google
lukaszkaiser@google.com

Slav Petrov Ilya Sutskever
Google Google
slav@google.com ilyasu@google.com

Geoffrey Hinton
Google
geoffhinton@google.com

https://arxiv.org/abs/1412.7449

deep

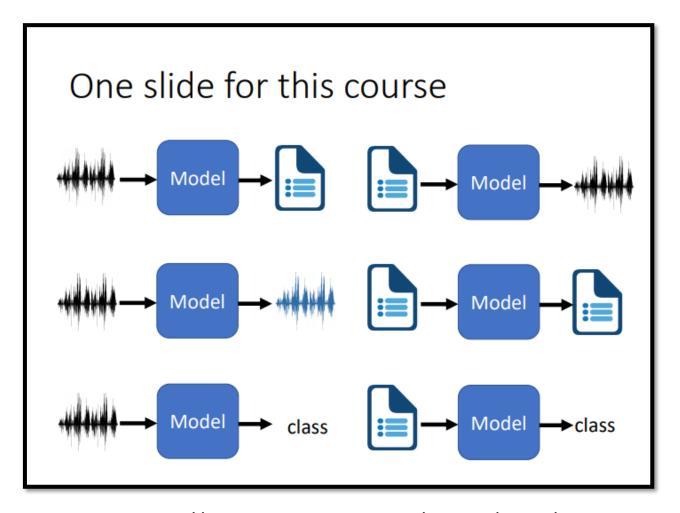
learning

is

very

powerful

Deep Learning for Human Language Processing 深度學習與人類語言處理

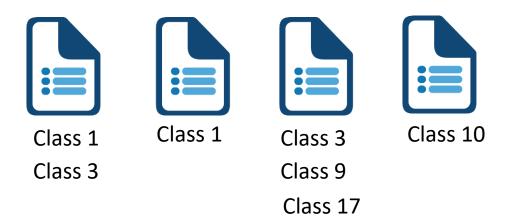


Source webpage: https://speech.ee.ntu.edu.tw/~hylee/dlhlp/2020-spring.html

c.f. Multi-class Classification

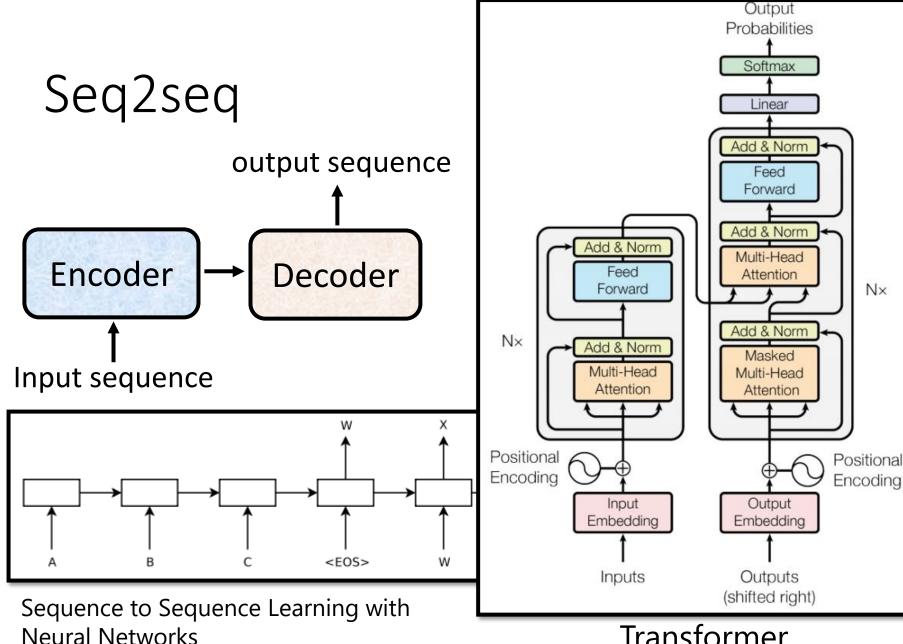
Seq2seq for Multi-label Classification

An object can belong to multiple classes.





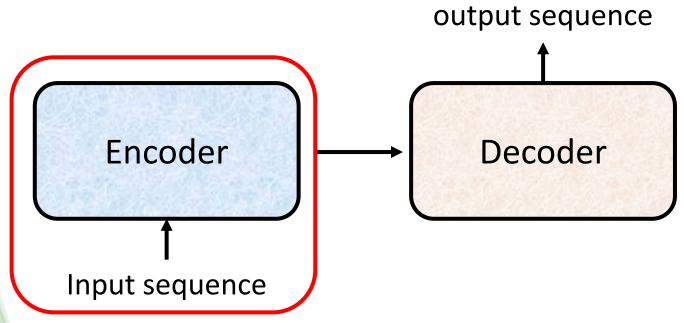
https://arxiv.org/abs/1909.03434 https://arxiv.org/abs/1707.05495



https://arxiv.org/abs/1409.3215

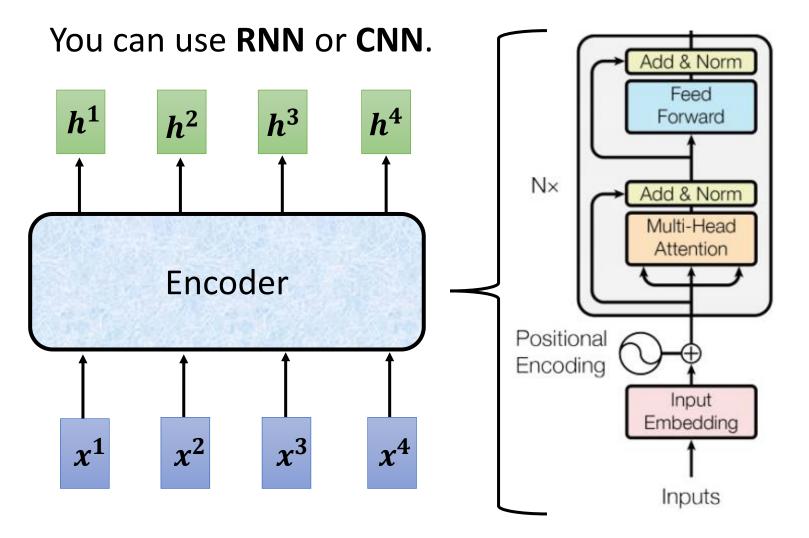
Transformer https://arxiv.org/abs/1706.03762 N×

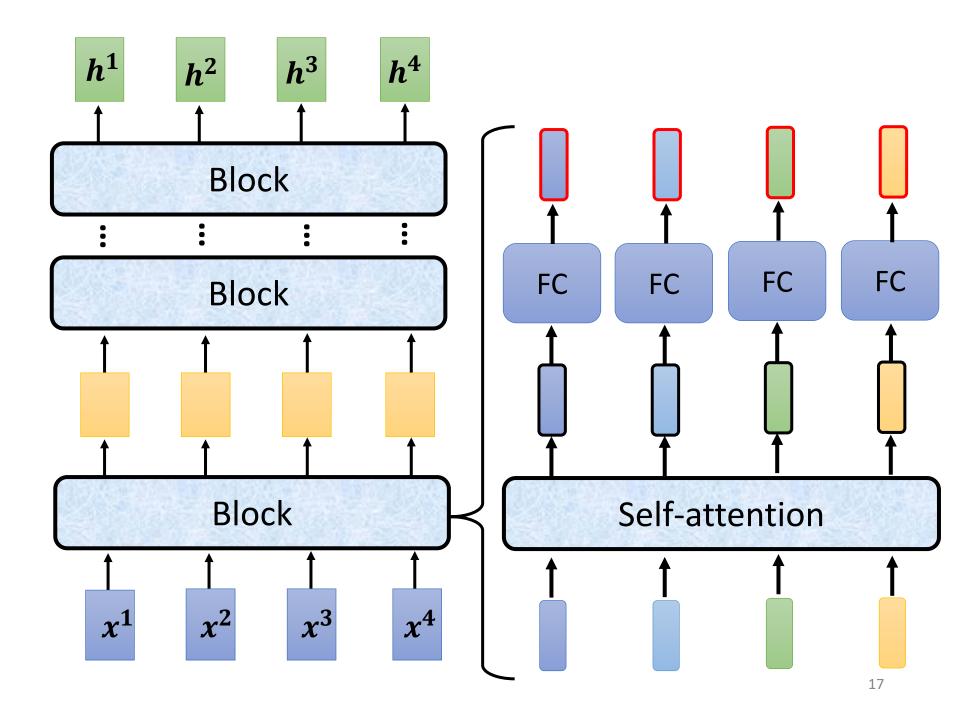
Encoder

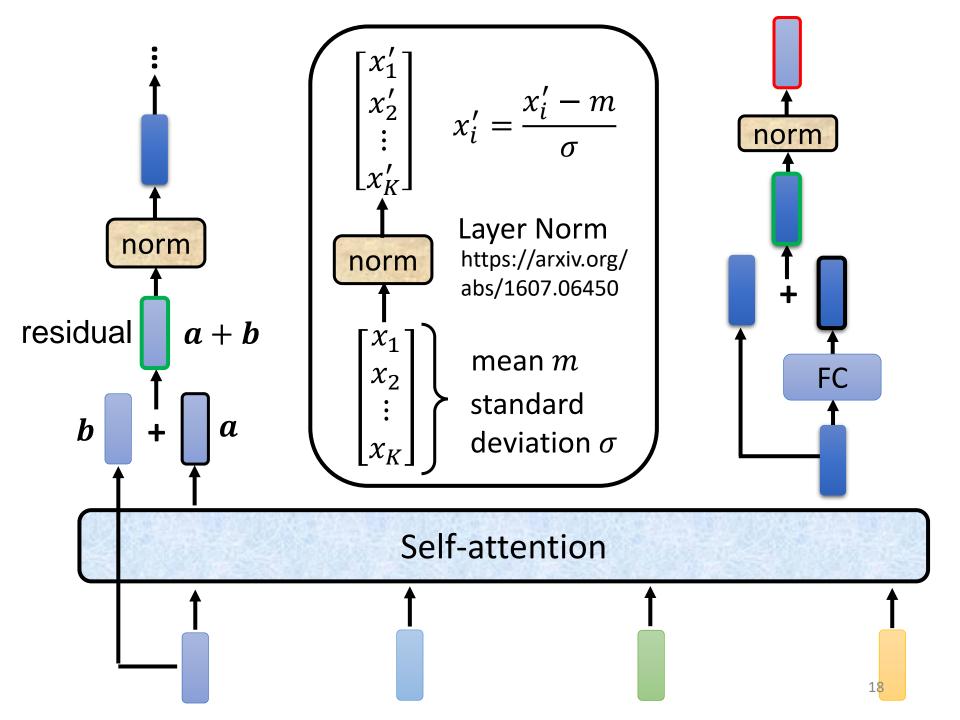


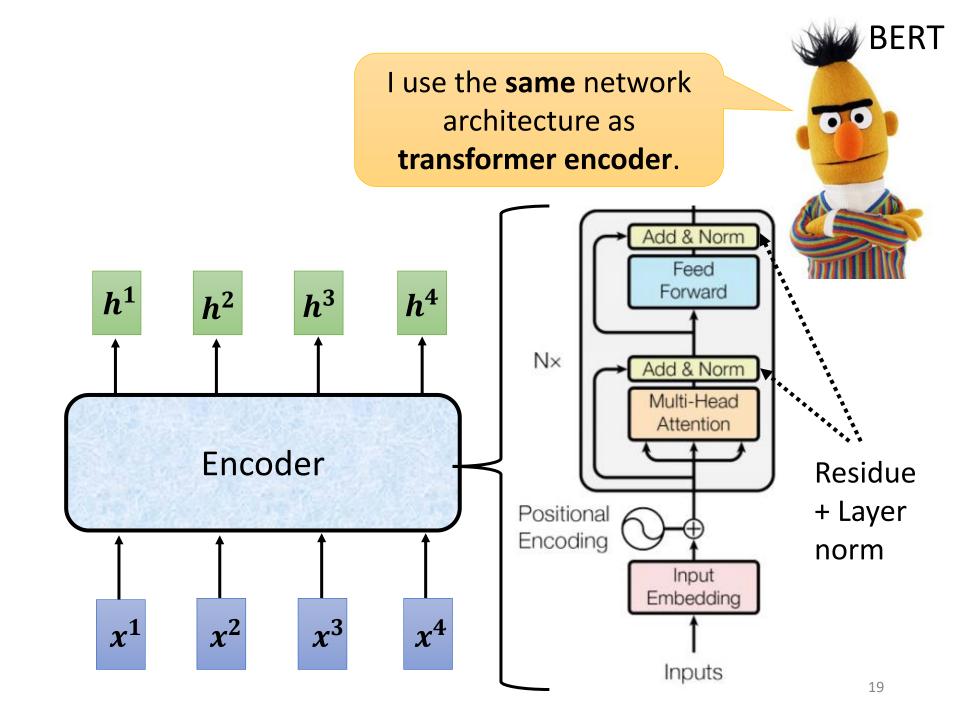
Encoder

Transformer's Encoder



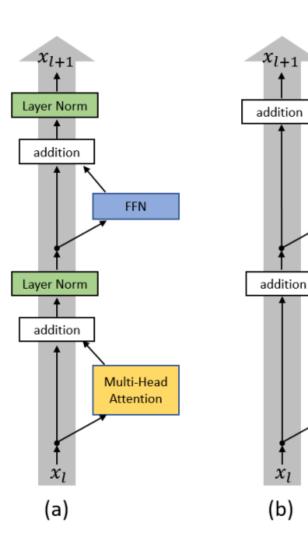






To learn more

- On Layer Normalization in the Transformer Architecture
- https://arxiv.org/abs/2002.047
 45
- PowerNorm: Rethinking Batch Normalization in Transformers
- https://arxiv.org/abs/2003.07845



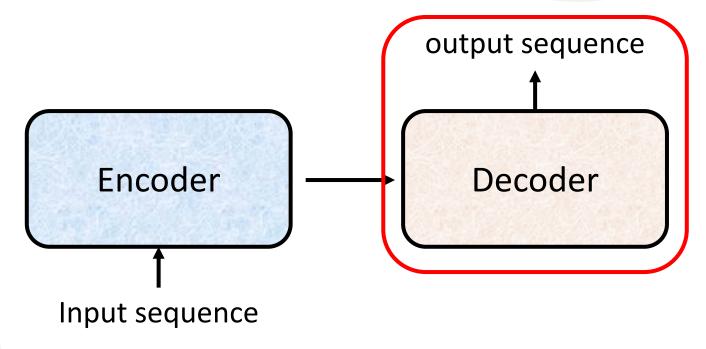
FFN

Layer Norm

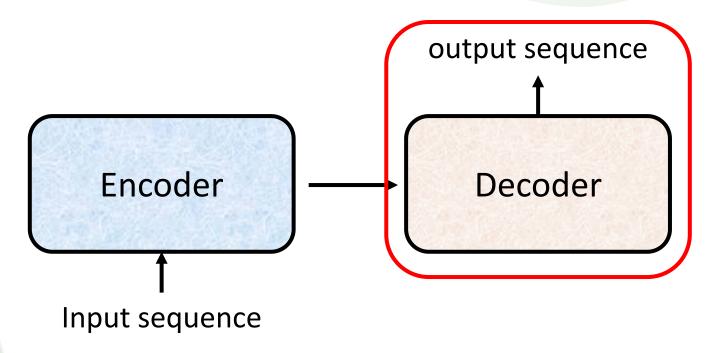
Multi-Head Attention

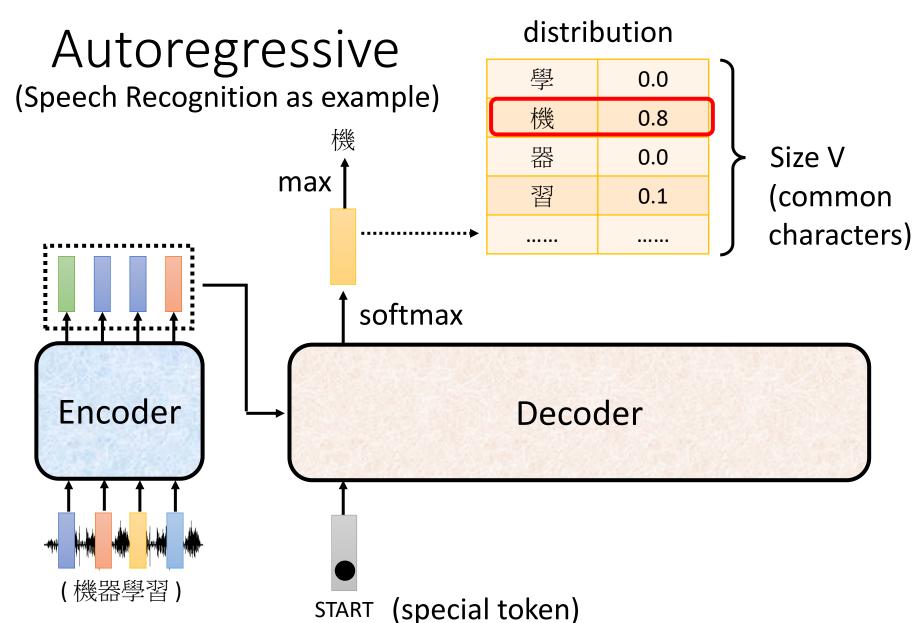
Layer Norm

Decoder

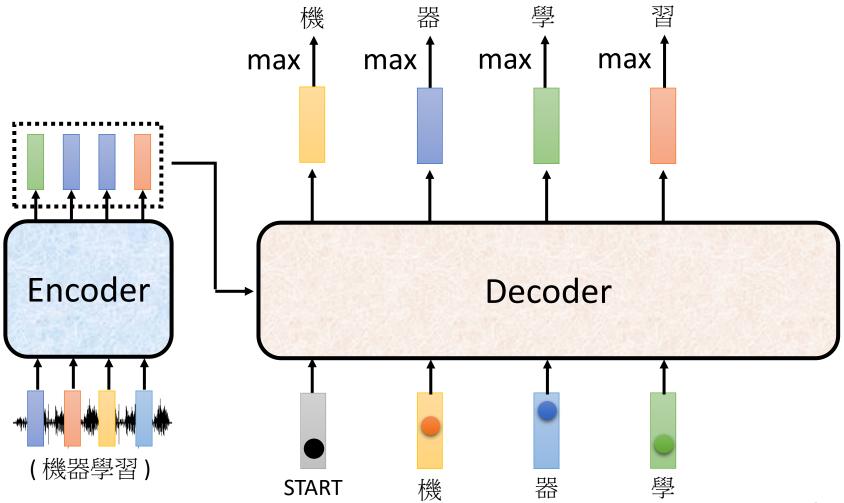


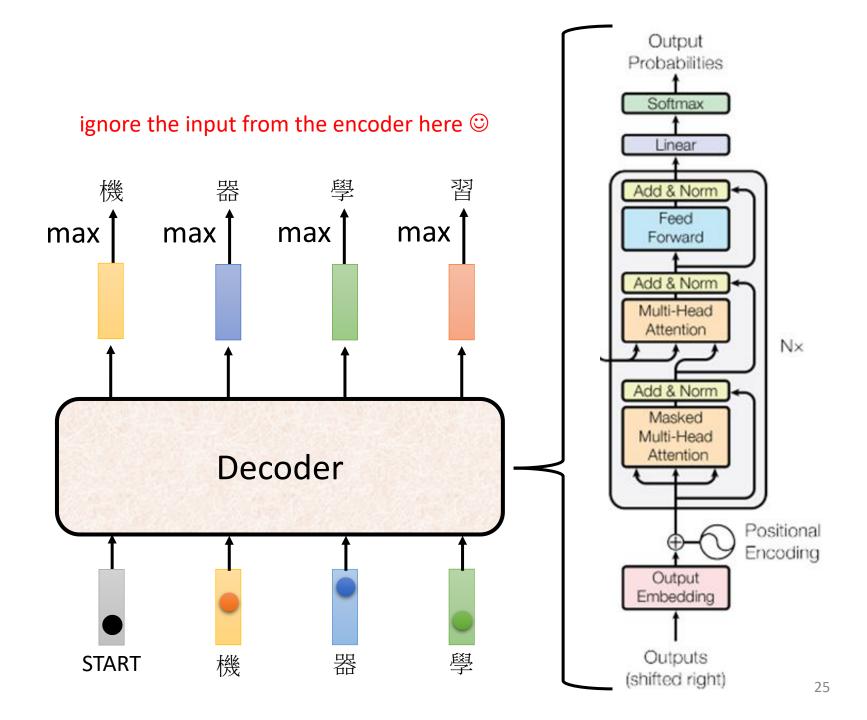
Decoder - Autoregressive (AT)

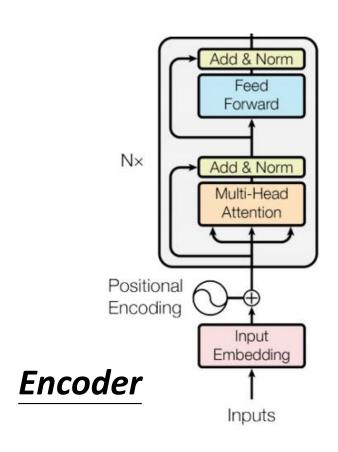


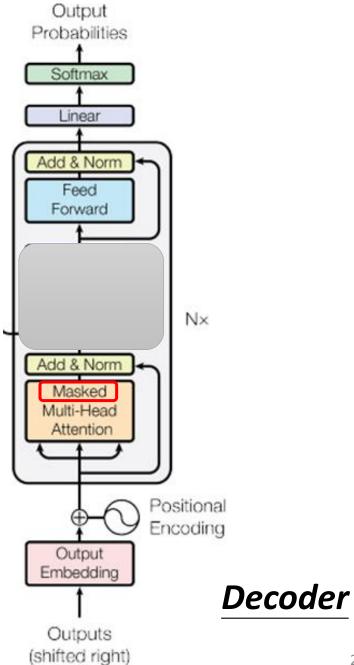


Autoregressive

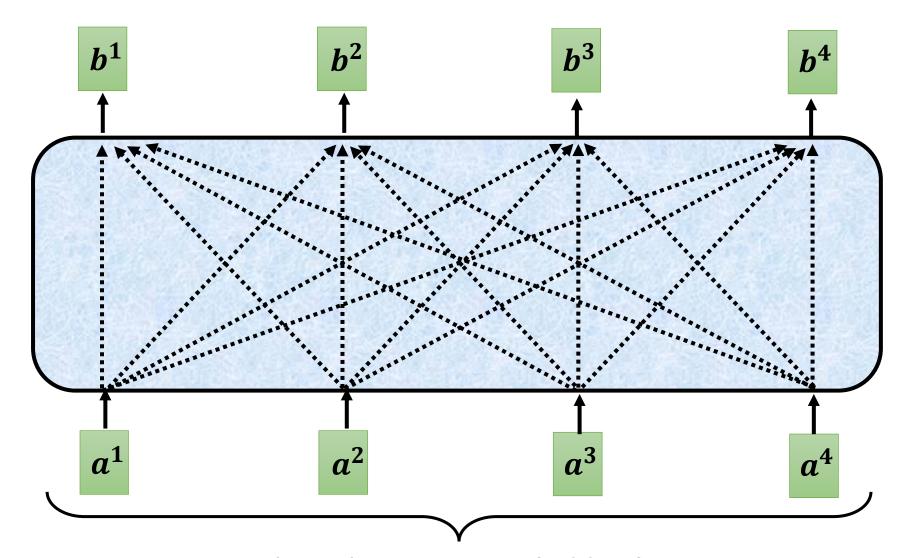






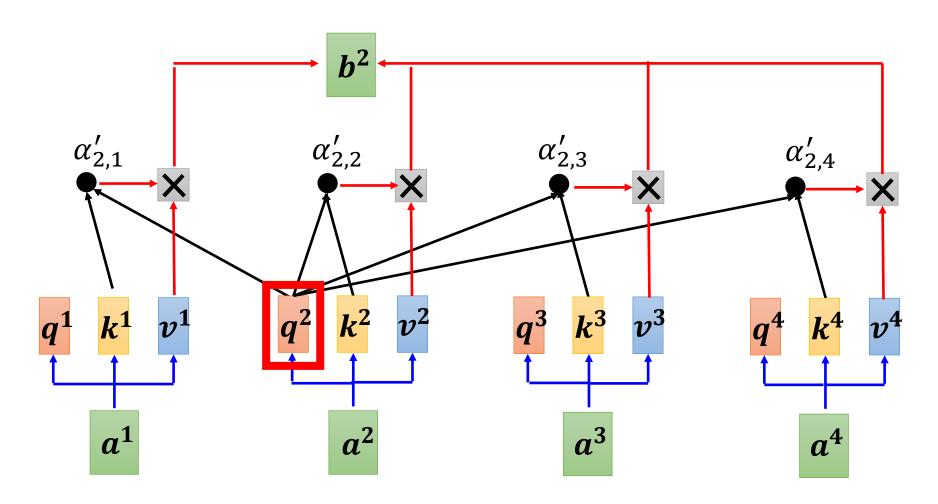


Self-attention → Masked Self-attention



Can be either input or a hidden layer

Self-attention → Masked Self-attention

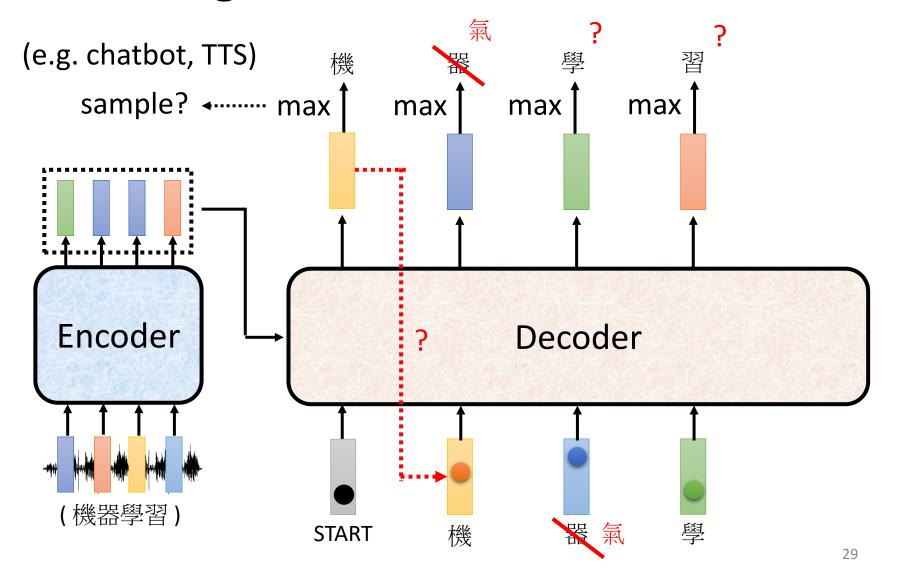


Why masked? Consider how does decoder work

Question? • Error propagation

Autoregressive

Distribution as input?



Autoregressive

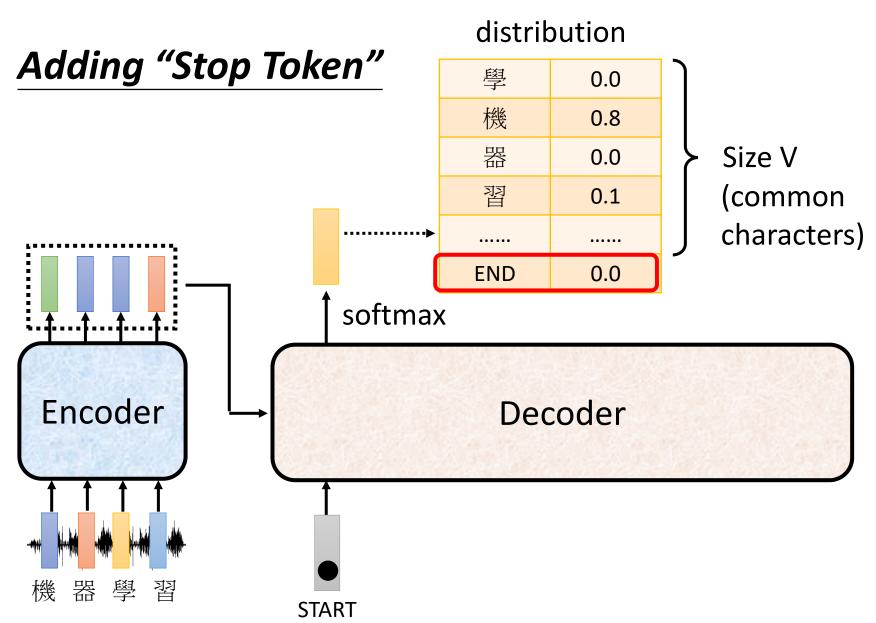
We do not know the correct output length.

30

Never stop! 器 習 機 max max max max max Encoder Decoder **START**

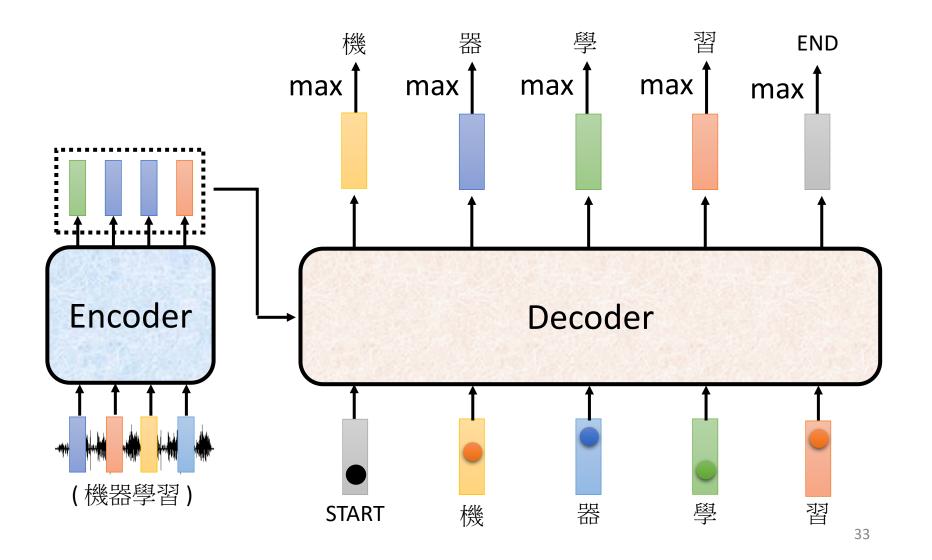
推文接龍 (Tweet Solitaire)

```
推
                 超
                                         06/12 10:39
推
                                         06/12 10:40
推
                                         06/12 10:41
          tion:
                                         06/12 10:47
          host:
推
                         中
                                         06/12 10:59
推
          403:
                                         06/12 11:11
推
                                         06/12 11:13
推
                                         06/12 11:17
                                         06/12 11:32
                                         06/12 12:15
推 tlkagk:
```

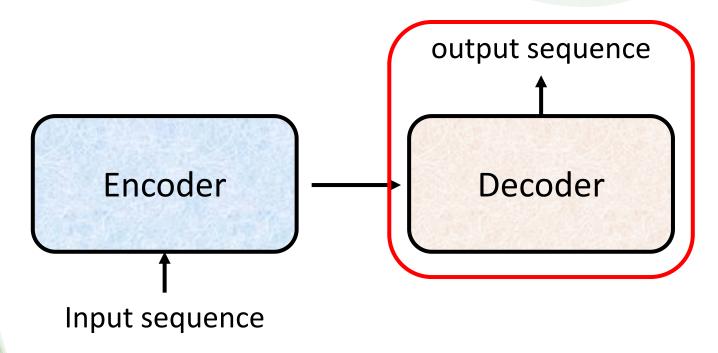


Autoregressive

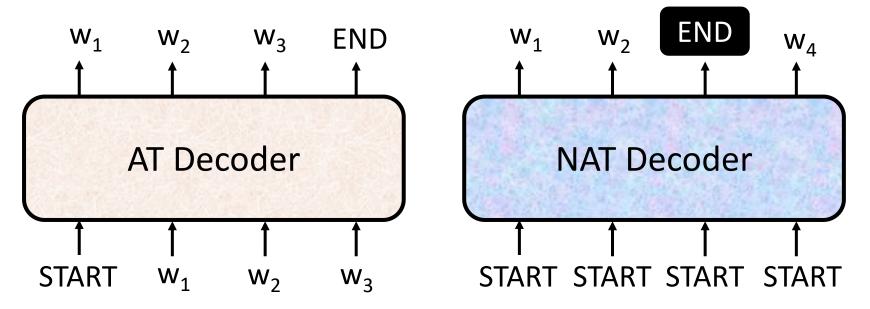
Stop at here!



Decoder - Non-autoregressive (NAT)



AT v.s. NAT



- ➤ How to decide the output length for NAT decoder?
 - Another predictor for output length
 - Output a very long sequence, ignore tokens after END
- > Advantage: parallel, more stable generation (e.g., TTS)
- ➤ NAT is usually worse than AT (why? Multi-modality)

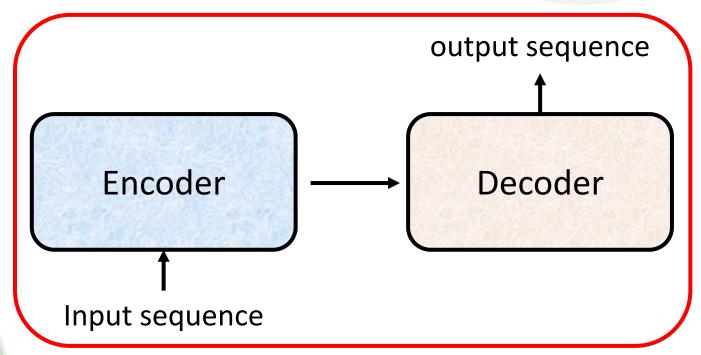
To learn more

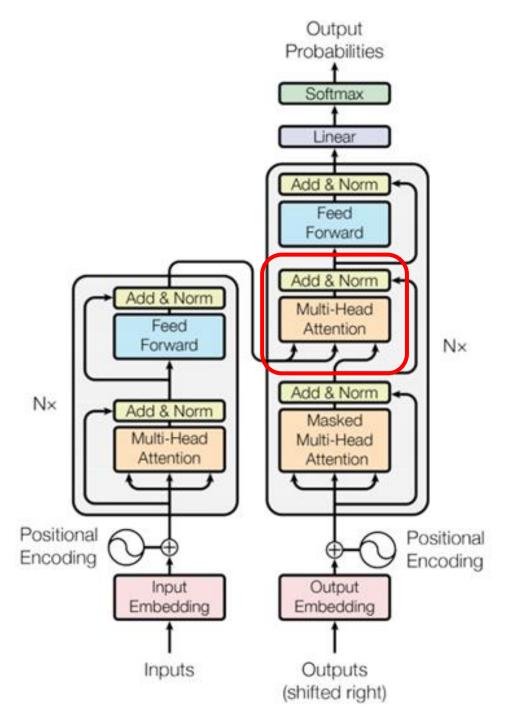


https://youtu.be/jvyKmU4OM3c (in Mandarin)



Encoder-Decoder





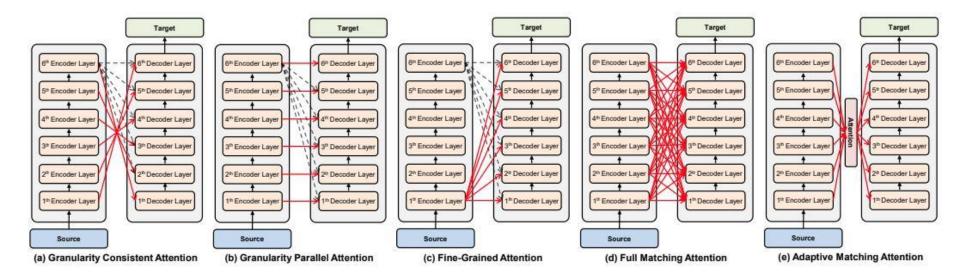


Figure 2: We present the proposal on Transformer with various strategies for routing the source representations: (a) Granularity Consistent Attention; (b) Granularity Parallel Attention; (c) Fine-Grained Attention; (d) Full Matching Attention; (e) Adaptive Matching Attention. The dashed lines represent the original attention to the last encoder layer and we omit them in (e) for clarity.

Training

