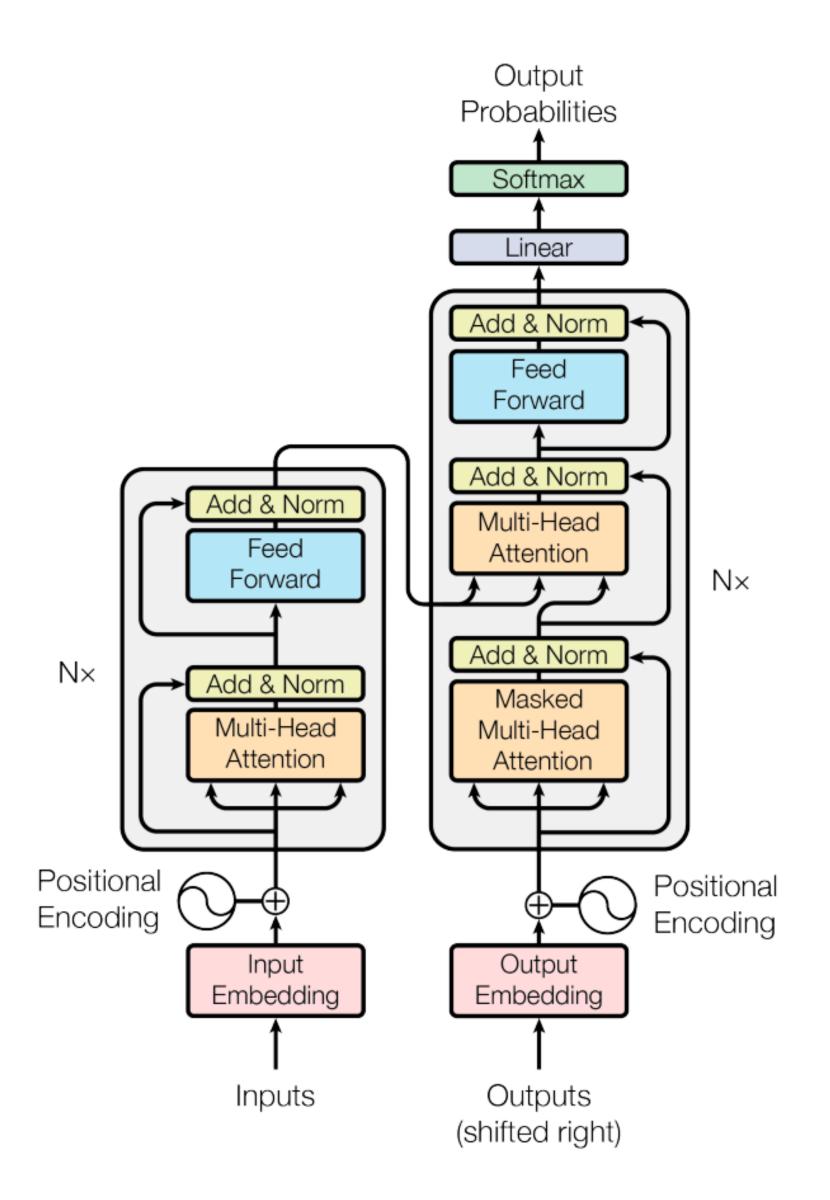
트랜스포머 최종 정리



이제 눈에 익숙한 우리 트랜스포머



Dec 2017

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Attention Is All You Need

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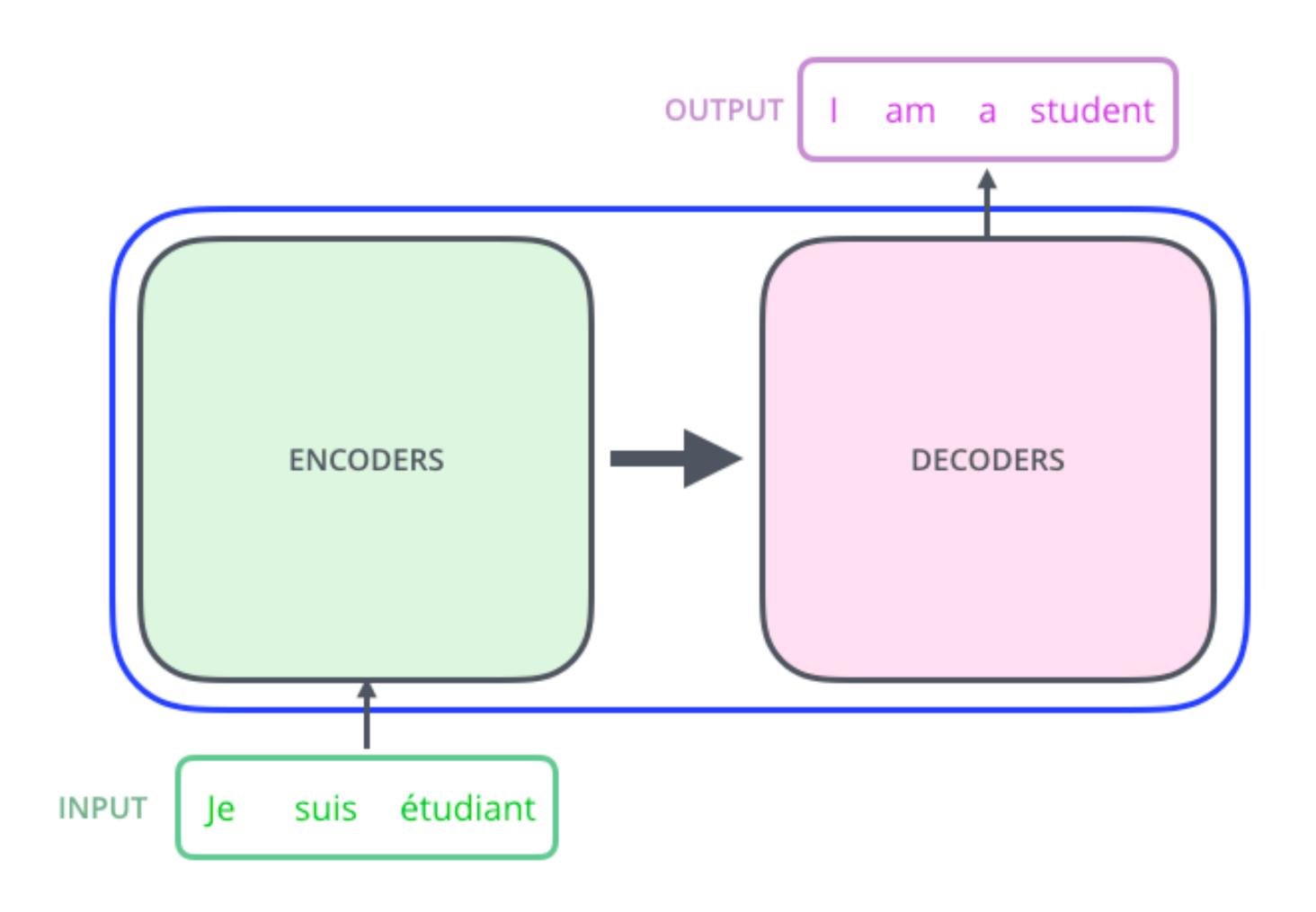
Illia Polosukhin* † illia.polosukhin@gmail.com





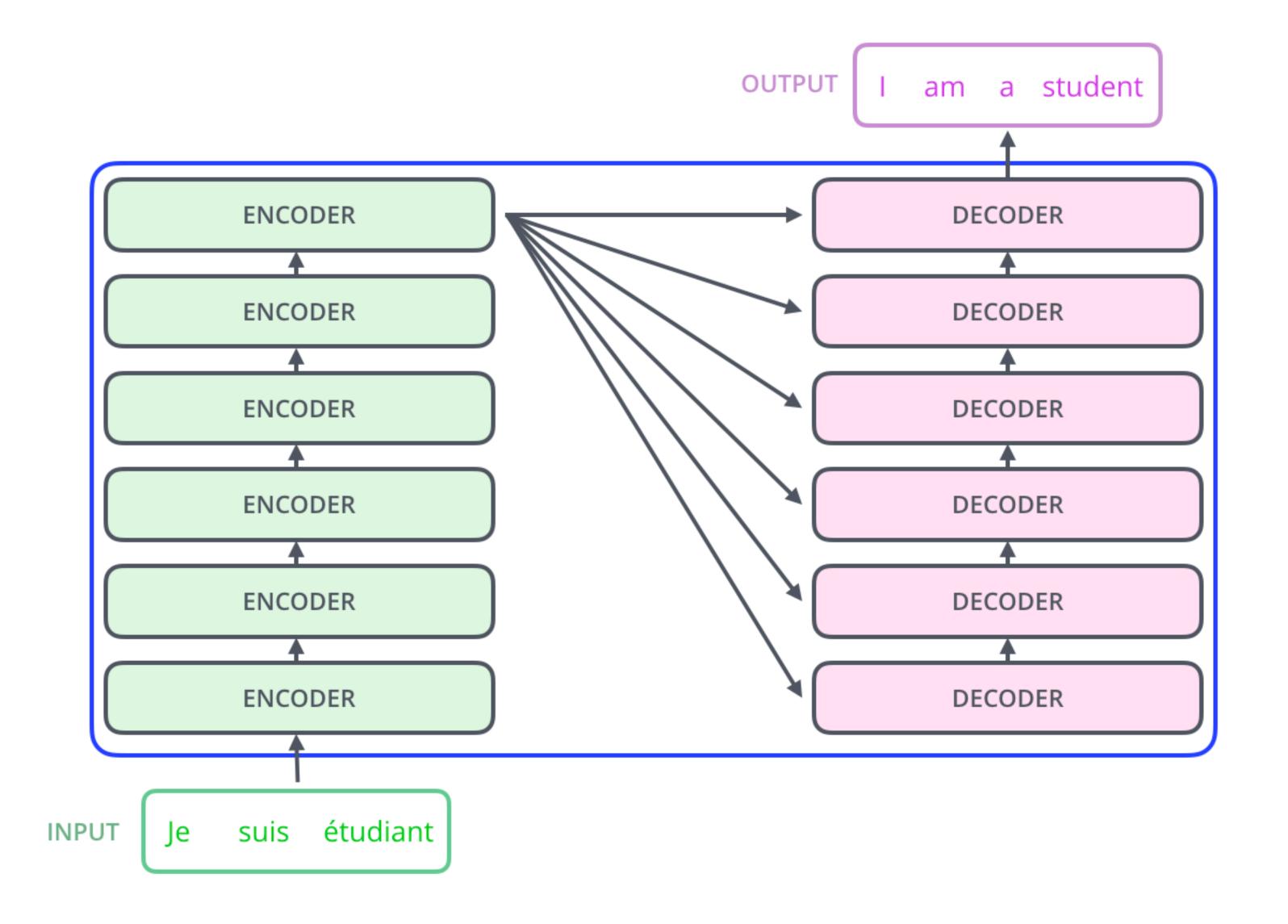
https://jalammar.github.io/illustrated-transformer/





https://jalammar.github.io/illustrated-transformer/



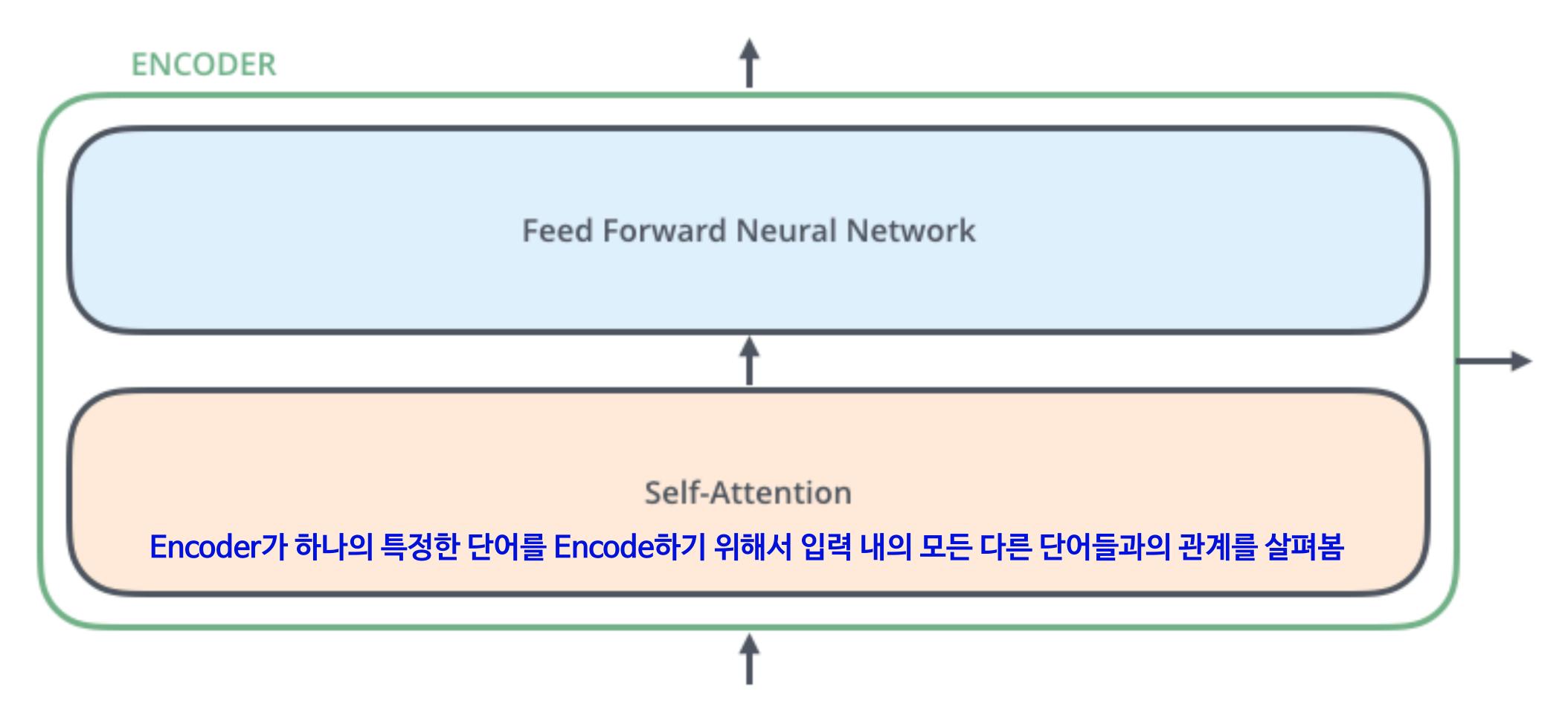


https://jalammar.github.io/illustrated-transformer/

https://nlpinkorean.github.io/illustrated-transformer/

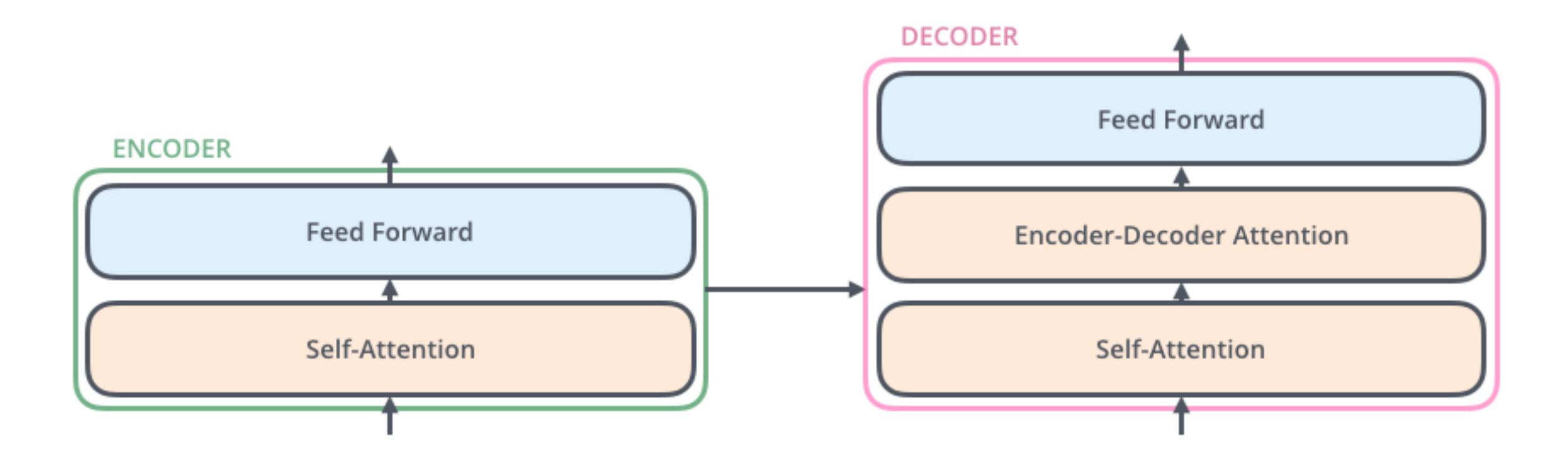


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https://jalammar.github.io/illustrated-transformer/





Encoder-Decoder Attention:
Decoder가 입력 문장 중에서 각 타임 스텝에서 가장 관련 있는 부분에 집중할 수 있도록 함

https://jalammar.github.io/illustrated-transformer/



트랜스포머의 입력을 다시 살펴 보겠습니다

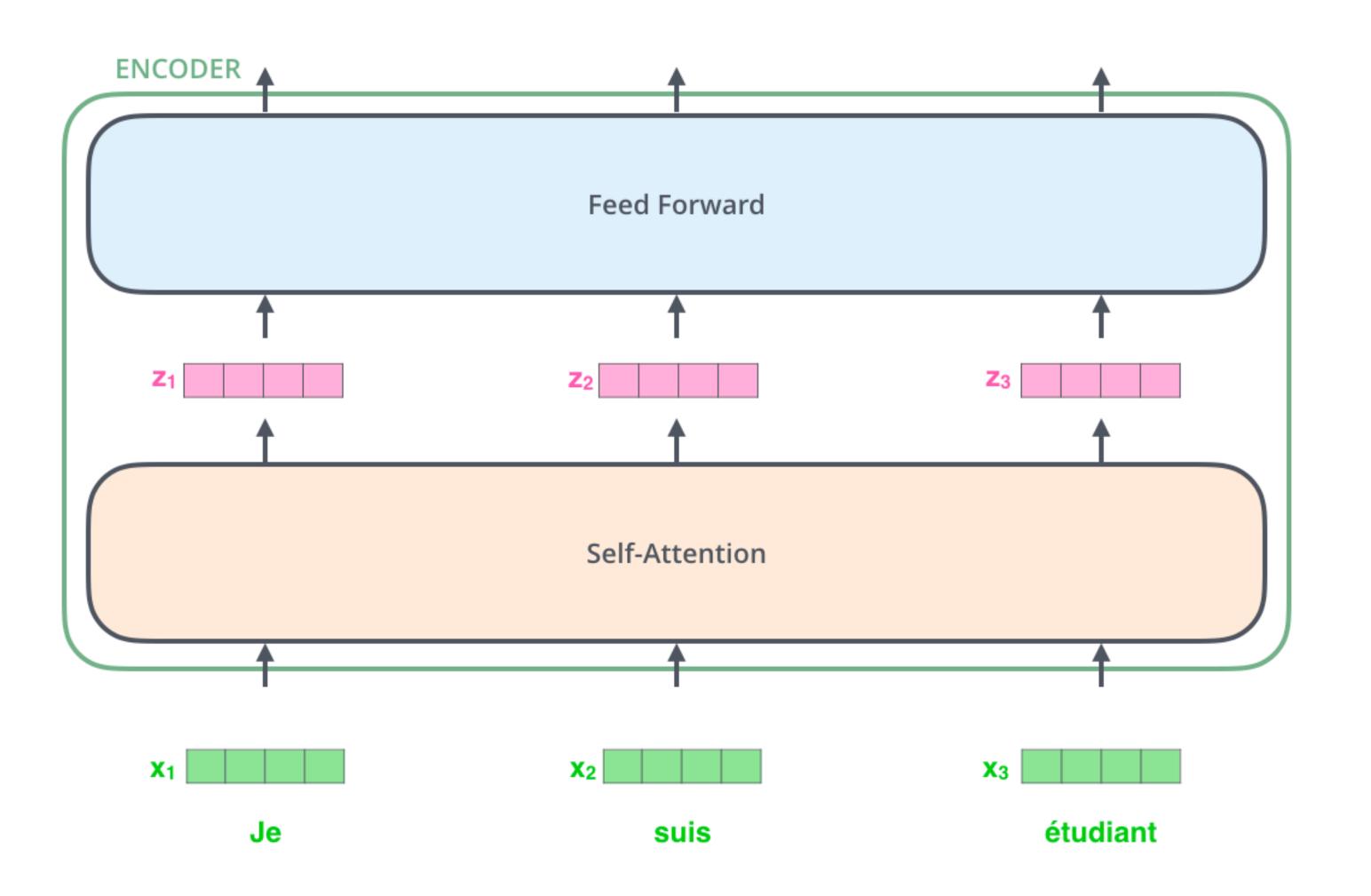


각 단어들을 크기 512의 벡터로 임베딩

https://jalammar.github.io/illustrated-transformer/



트랜스포머의 인코더를 거치며 self-encoding 과정을 거칩니다



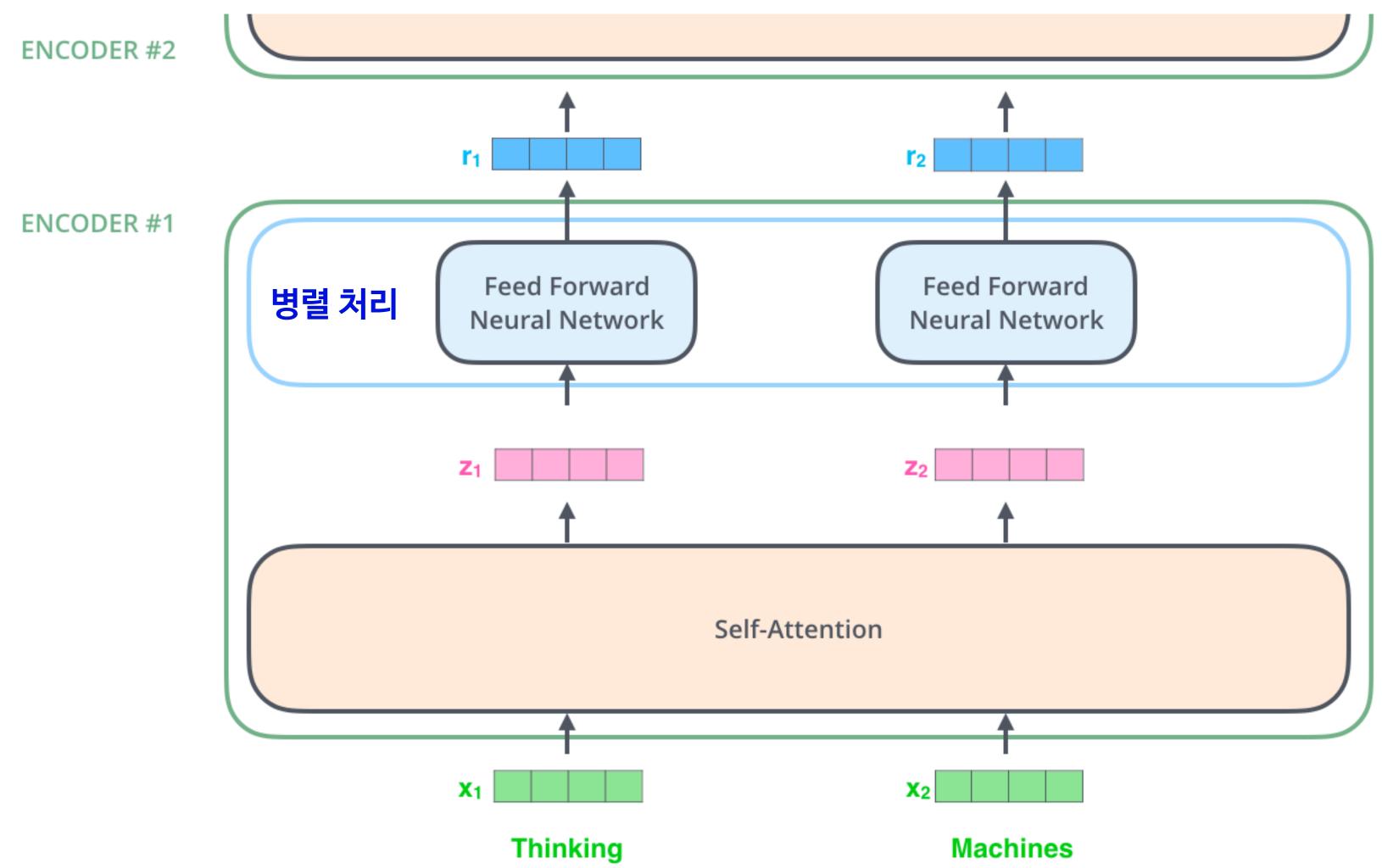
https://jalammar.github.io/illustrated-transformer/

https://nlpinkorean.github.io/illustrated-transformer/



https://core.today

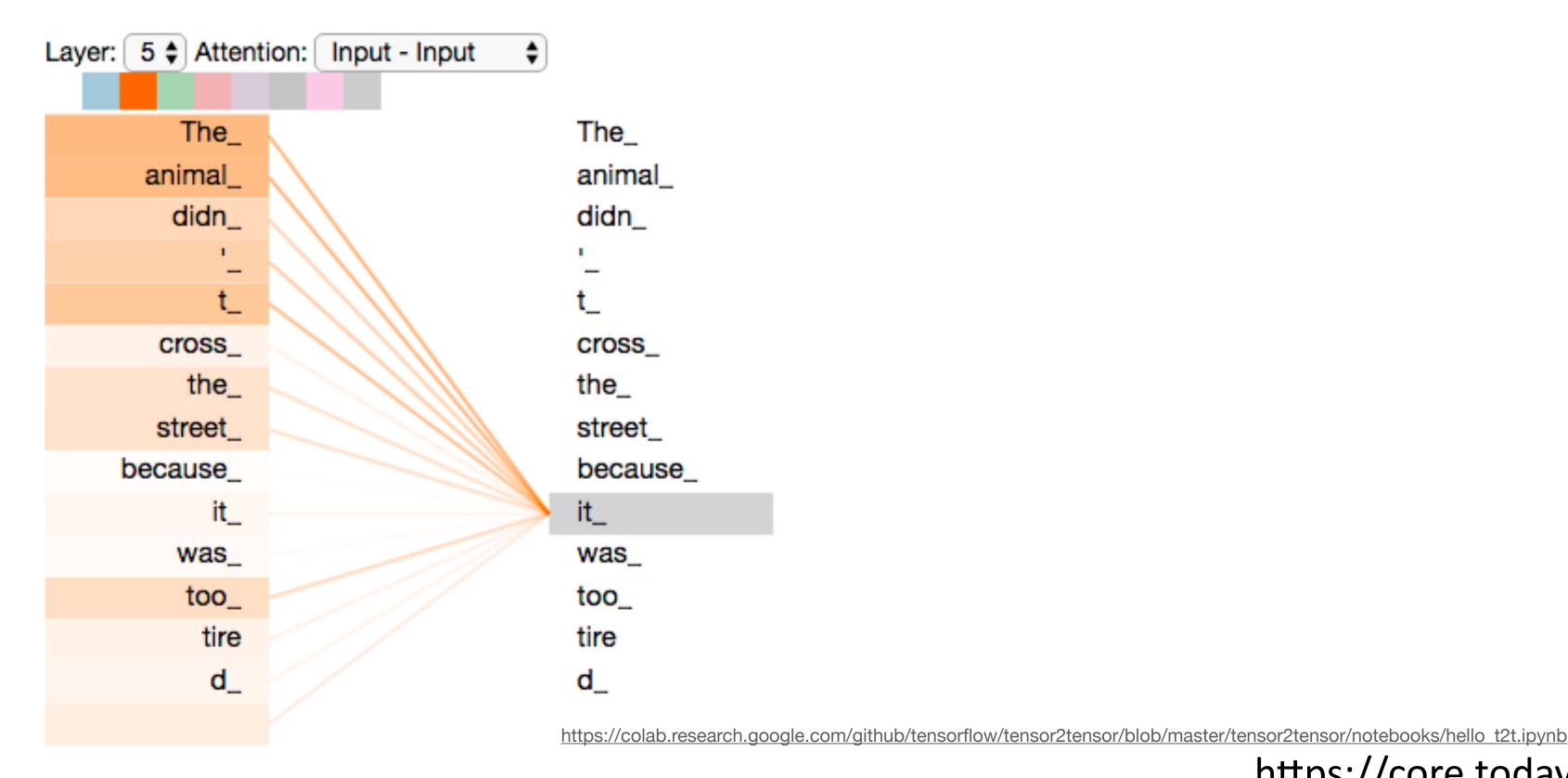
FFNN은 의존성이 없어 병렬 처리가 가능합니다



https://jalammar.github.io/illustrated-transformer/

Self-Attention은 무엇이냐구요?

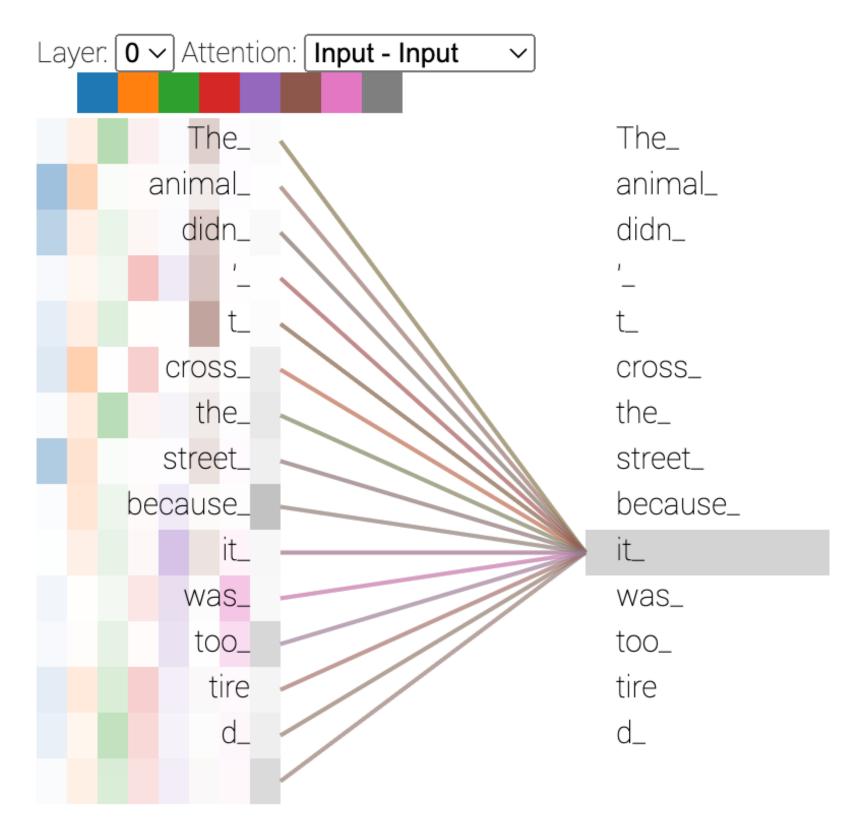
The animal didn't cross the street because it was too tired





Self-Attention은 무엇이냐구요?

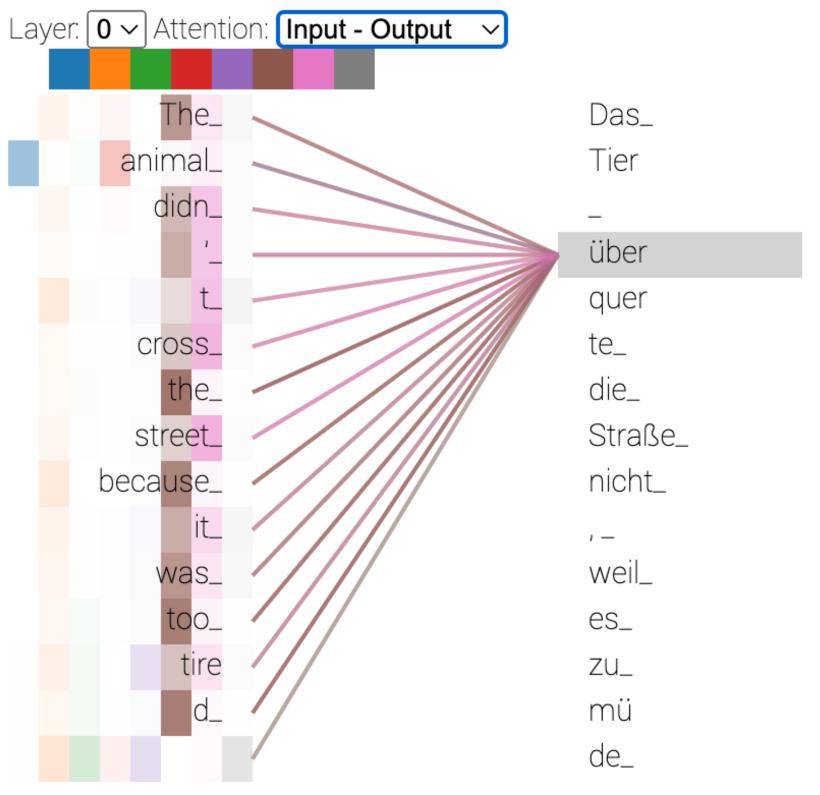
The animal didn't cross the street because it was too tired





Self-Attention은 무엇이냐구요?

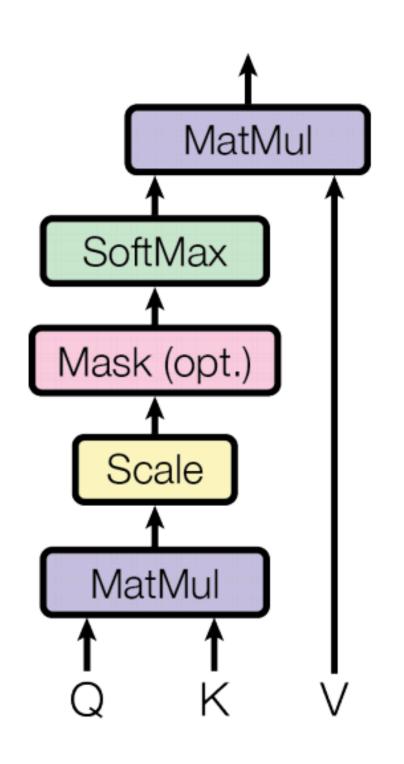
The animal didn't cross the street because it was too tired

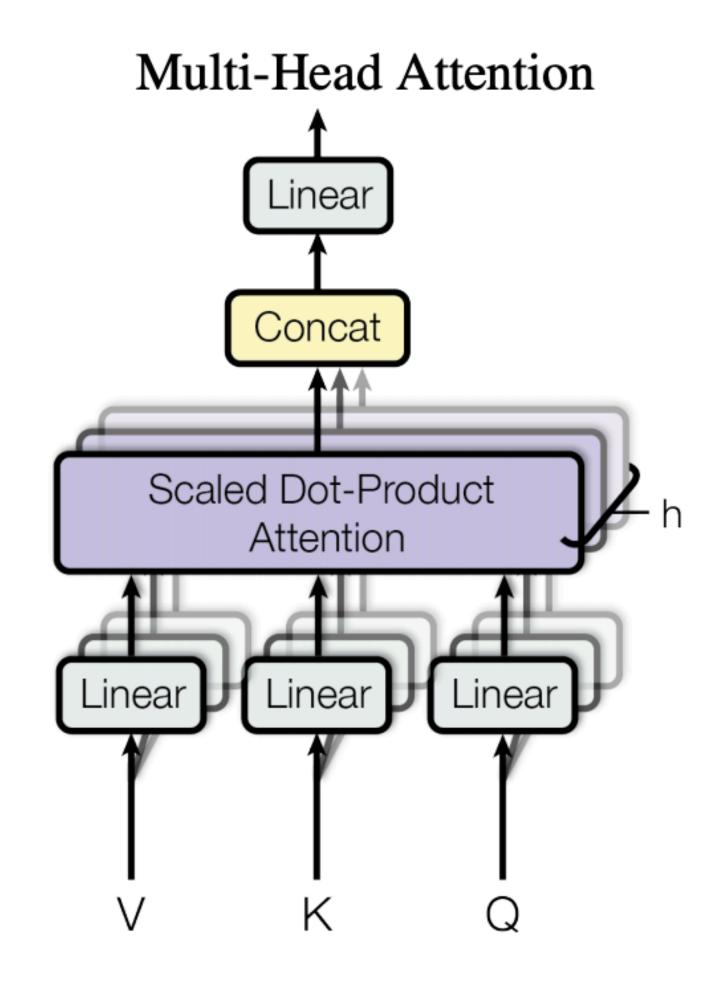




멀티-헤드 어텐션

Scaled Dot-Product Attention

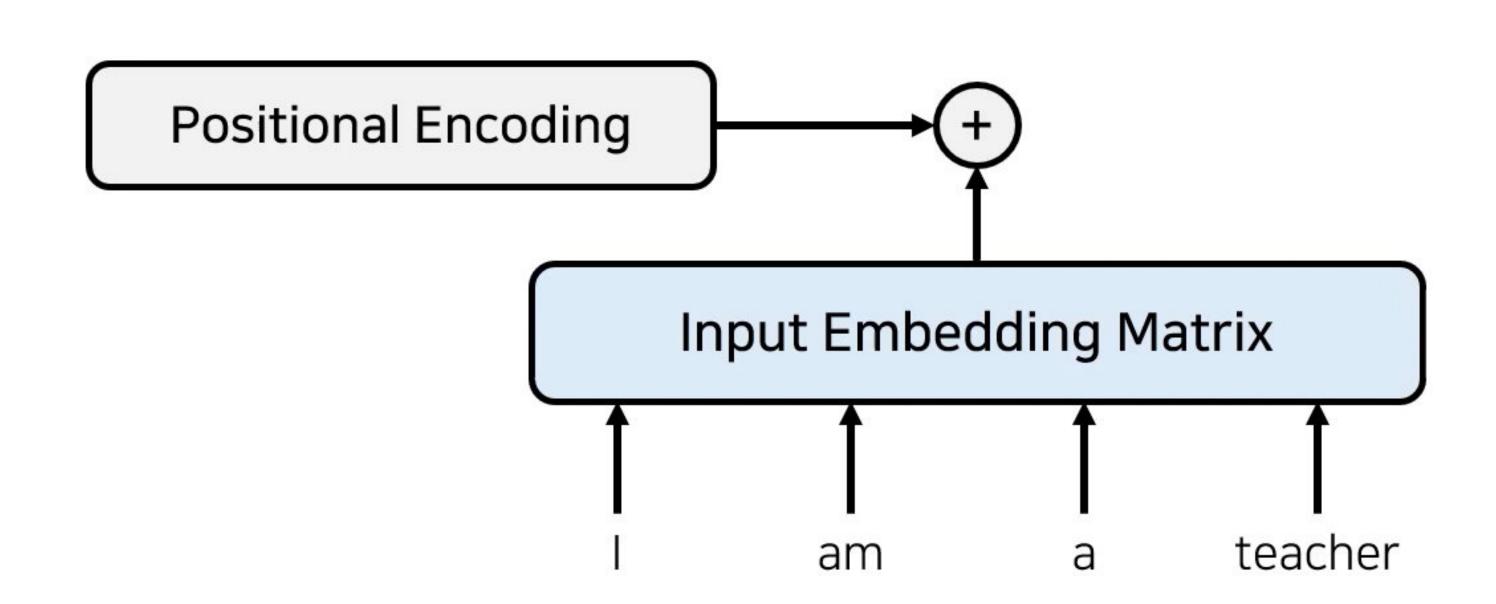






위치 인코딩으로 단어의 순서 정보를 반영합니다

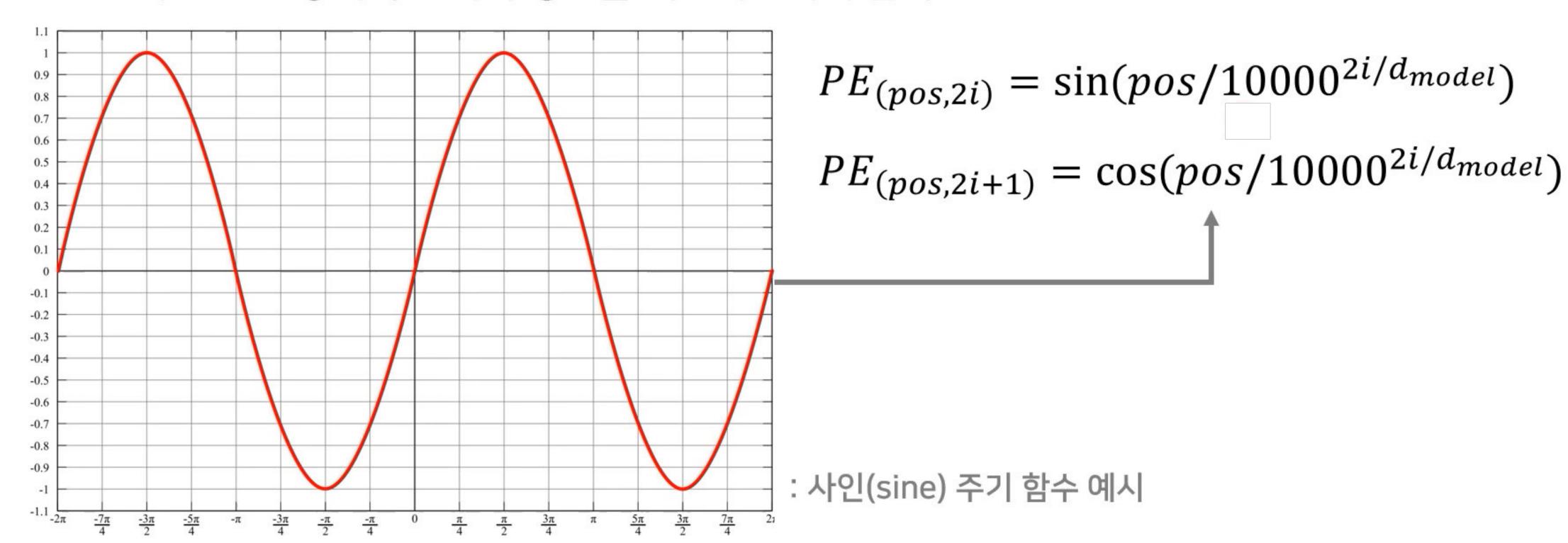
- RNN을 사용하지 않으려면 위치 정보를 포함하고 있는 임베딩을 사용해야 합니다.
 - 이를 위해 트랜스포머에서는 Positional Encoding을 사용합니다.





위치 정보는 sin과 cos으로 인코딩

- Positional Encoding은 다음과 같이 주기 함수를 활용한 공식을 사용합니다.
 - 각 단어의 상대적인 위치 정보를 네트워크에게 입력합니다.

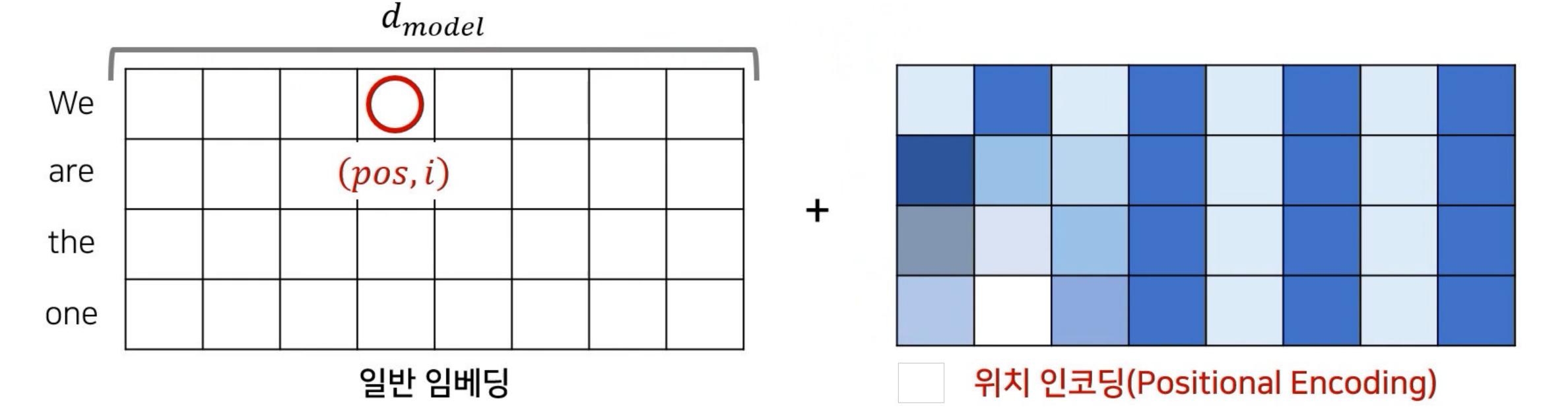




• Positional Encoding은 다음과 같이 주기 함수를 활용한 공식을 사용합니다.

$$PE_{(pos,2i)} = \sin(pos/10000^{2i/d_{model}})$$

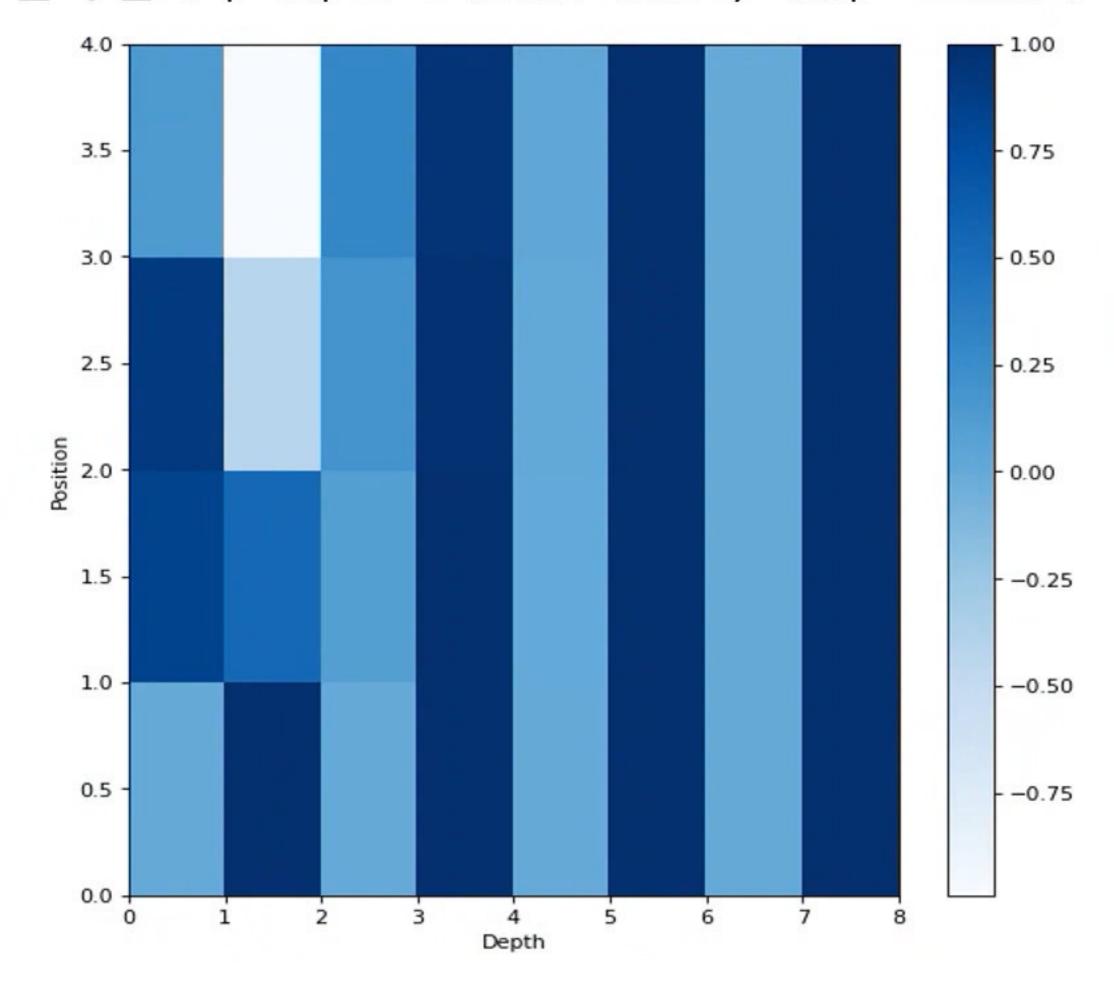
$$PE_{(pos,2i+1)} = \cos(pos/10000^{2i/d_{model}})$$





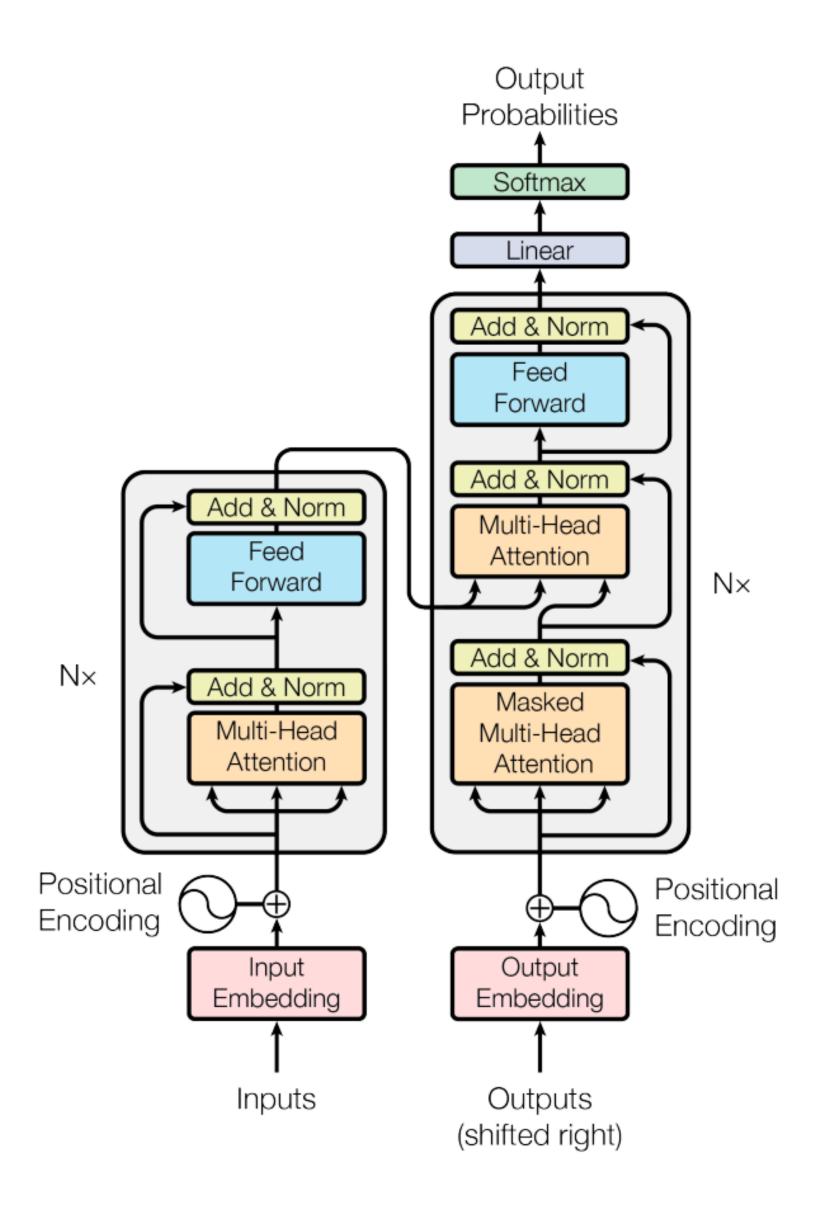
```
import math
import matplotlib.pyplot as plt
n = 4 # 단어(word)의 개수
dim = 8 # 임베딩(embedding) 차원
def get_angles(pos, i, dim):
    angles = 1 / \text{math.pow}(10000, (2 * (i // 2)) / \text{dim})
   return pos * angles
def get_positional_encoding(pos, i, dim):
   if i % 2 == 0: # 짝수인 경우 사인 함수
       return math.sin(get_angles(pos, i, dim))
   # 홀수인 경우 코사인 함수
   return math.cos(get_angles(pos, i, dim))
result = [[0] * dim for _ in range(n)]
for i in range(n):
    for j in range(dim):
       result[i][j] = get_positional_encoding(i, j, dim)
```

출력 결과: plt.pcolormesh(result, cmap='Blues')





트랜스포머 끝!!



Attention Is All You Need

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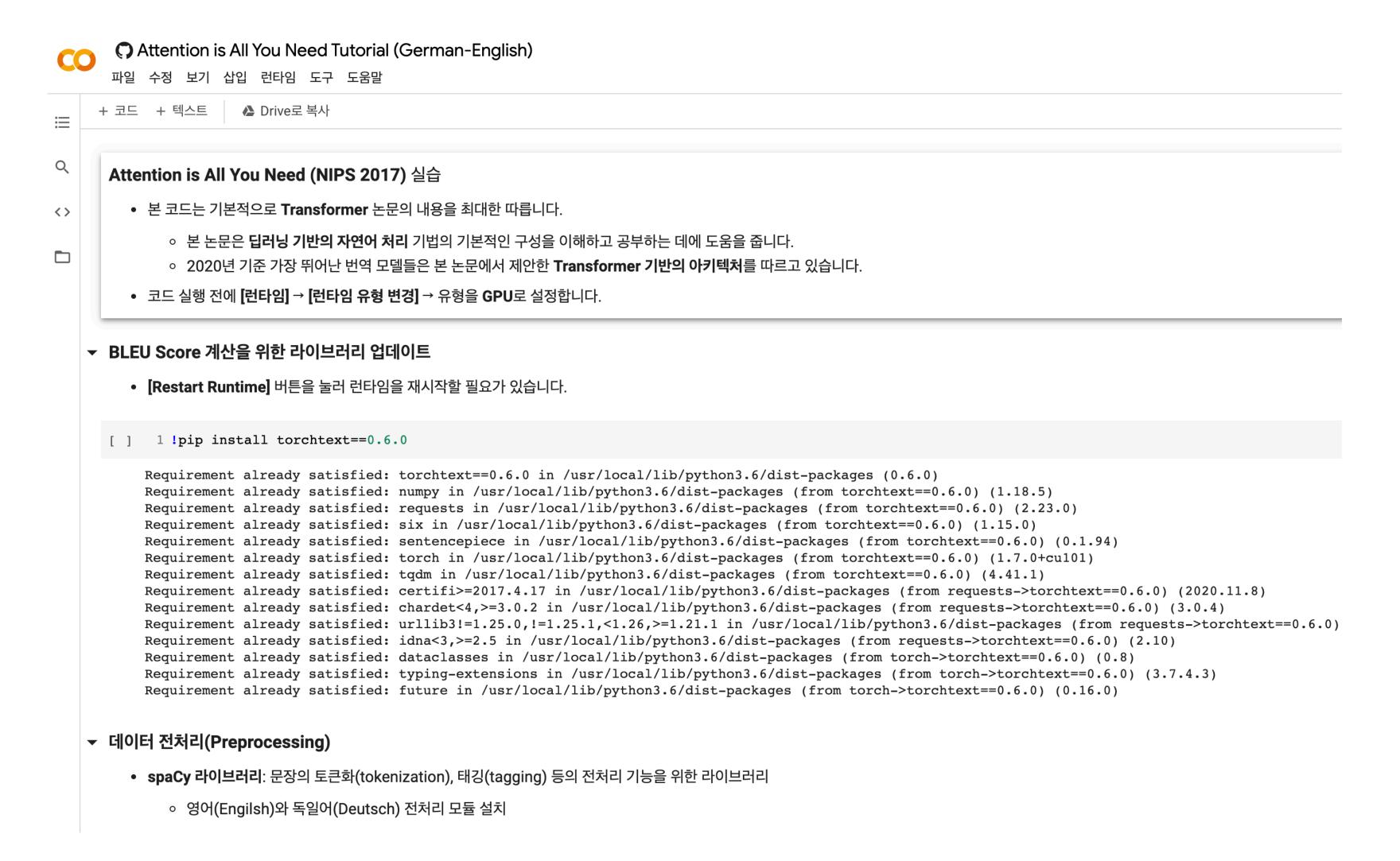
https://arxiv.org/pdf/1706.03762.pdf

https://core.today

Dec

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Attention is All You Need 실습



https://colab.research.google.com/github/ndb796/Deep-Learning-Paper-Review-and-Practice/blob/master/code_practices/



Attention is All You Need 실습

```
Epoch: 08 | Train Loss: 1.921 | Time: 698.7559933662415
       40% ■■■■
epoch:
                       | 8/20 [1:39:53<2:34:40, 773.41s/it]
저장한 파일 이름 : ./epoch 7.pth 저장하는 데 걸린 시간 : 2.6485018730163574
                          0/6250 [00:00<?, ?it/s]
           0 % |
training:
           1%||
                          88/6250 [00:10<11:43, 8.76it/s]
training:
           3%||
                          176/6250 [00:20<11:41, 8.66it/s]
training:
           3%||
training:
                          176/6250 [00:30<11:41, 8.66it/s]
           4 % |
                          255/6250 [00:30<12:09, 8.22it/s]
training:
           5%
training:
                          338/6250 [00:40<11:59, 8.22it/s]
           7% 📗
                          428/6250 [00:50<11:26, 8.49it/s]
training:
          8 8 📗
                          518/6250 [01:01<11:12, 8.52it/s]
training:
training:
         10%
                          604/6250 [01:11<11:13, 8.38it/s]
```

https://github.com/CoreDotToday/DeepLearningTextBook/blob/main/NLP/Transformer/Attention_is_All_You_Need%EC%9D%98_%EC%B5%9C%EC%A2%85%EB%B3%B8.ipynb



Attention is All You Need 실습

```
# 모델을 평가 모드로 설정
model.eval()
# 입력 문장
sent = '나'
# 문장 토큰화
proc_sent = kor_tokenizer.encode(sent)
# 후처리
post_proc_sent = postprocess(proc_sent.ids)
# Tensor로 변환 및 배치 차원 추가
input_tensor = torch.LongTensor(post_proc_sent).to(device)
input_tensor = input_tensor.unsqueeze(0)
# 출력 문장 초기화
output_tensor = torch.LongTensor([1]).to(device).unsqueeze(0) # <sos> 토큰 + 배치 차원 추가
# 디코딩
with torch.no_grad():
   for _ in range(50): # 최대 길이
       logits, _ = model(input_tensor, output_tensor)
       next_token = logits.argmax(-1)[:,-1]
       output_tensor = torch.cat([output_tensor, next_token.unsqueeze(-1)], dim=-1)
       if next_token.item() == 2: # <eos> 星己
           break
# 결과 디코딩
decoded_output = eng_tokenizer.decode(output_tensor.squeeze().tolist())
print(decoded_output)
```

